Functional Annotation of Human Long Non-Coding RNAs via Molecular Phenotyping

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Abstract

Long non-coding RNAs (lncRNAs) constitute the majority of transcripts in mammalian genomes and yet, their functions remain largely unknown. We systematically suppressed 285 lncRNAs in human dermal fibroblasts and quantified cellular growth, morphological changes, and transcriptomic responses using Capped Analysis of Gene Expression (CAGE). The resulting transcriptomic profiles recapitulated the observed cellular phenotypes, yielding specific roles for over 40% of analyzed lncRNAs in regulating distinct biological pathways, transcriptional machinery, alternative promoter activity and architecture usage. Overall, combining cellular and molecular profiling provided a powerful approach to unravel the distinct functions of lncRNAs, which we highlight with specific functional roles for *ZNF213-AS1* and *lnc-KHDC3L-2.*

Introduction

Over 50,000 loci in the human genome transcribe long non-coding RNA (lncRNA) (Hon et al. 2017; Iyer et al. 2015), which are defined as transcripts at least 200 nt long with low or no protein-coding potential. While lncRNA genes outnumber protein-coding genes in mammalian genomes, they are comparatively less conserved (Ulitsky 2016), lowly expressed, and more cell-type-specific (Hon et al. 2017). However, the evolutionary conservation of lncRNA promoters (Carninci et al. 2005) and their structural motifs (Derrien et al. 2012) suggest that lncRNAs are fundamental biological regulators. To date, only a few hundred human lncRNAs have been extensively characterized (Quek et al. 2015; Volders et al. 2015; de Hoon et al. 2015), revealing their roles in regulating transcription (Engreitz, Ollikainen, et al. 2016), translation (Carrieri et al. 2012), and chromatin state (Gupta et al. 2010; Guttman and Rinn 2012; Guttman et al. 2011); (Ransohoff et al. 2018; Quinn and Chang 2016).

Our recent FANTOM5 analysis showed that 19,175 (out of ~27,000) human lncRNA loci are functionally implicated (Hon et al. 2017). However, genomic screens are necessary to comprehensively characterize each lncRNA. One common approach of gene knockdown followed by a cellular phenotype assay typically characterizes a small percentage of lncRNAs for a single observable phenotype. For example, a recent large-scale screening using CRISPR interference (CRISPRi) found that approximately 500 lncRNA loci (~3.7% of targeted lncRNA loci) are essential for cell growth or viability in a cell-type specific manner (Liu et al. 2017). In addition, CRISPR-Cas9 experiments targeting splice sites identified ~2.1% of lncRNAs that affect growth

of K562 chronic myeloid leukemia cells (Liu et al. 2018). A CRISPR activation study has also

revealed ~0.11% lncRNAs to be important for drug resistance in melanoma (Joung et al. 2017).

As a part of the FANTOM 6 pilot project, we established an automated high-throughput cell culture

platform to suppress 285 lncRNAs expressed in human primary dermal fibroblasts (HDF) using

antisense LNA-modified gapmer antisense oligonucleotide (ASO) technology (Roux et al. 2017).

We then quantified the effect of each knockdown on cell growth and morphology using real-time

imaging, followed by Cap Analysis Gene Expression (CAGE; Murata et al. 2014) deep sequencing

to reveal molecular pathways associated with each lncRNA. In contrast to cellular phenotyping,

molecular phenotyping provides a detailed assessment of the response to an lncRNA knockdown

at the molecular level, allowing biological functions to be associated to lncRNAs even in the

absence of an observable cellular phenotype. We further investigated changes in promoter usage

and architecture modulated by the knockdowns, and integrated genome-wide chromosome

conformation capture data (Hi-C) with the CAGE data to explore regulatory functions of the

lncRNA targets. All data produced for this study are publicly available at

http://fantom.gsc.riken.jp/6/datafiles (user: review; password: f6plj0206) and analysis results can

be explored at http://fantom.gsc.riken.jp/zenbu/reports/#FANTOM6

115 Results

116 Selection and ASO-mediated knockdown of lncRNA targets

Human dermal fibroblasts (HDF) are non-transformed primary cells that are commonly used for

investigating cellular reprogramming (Takahashi et al. 2007; Ambasudhan et al. 2011), wound-

healing (Li and Wang 2011), fibrosis (Kendall R., et al 2014), and cancer (Kalluri 2016).

Fibroblasts are easy to transfect, proliferate in culture, and exhibit a distinctive spindle-shaped

morphology indicative of their mesenchymal origin, making it an ideal model for cellular

phenotypic screenings. Here, an unbiased selection of lncRNAs expressed in HDF was performed

to choose 285 lncRNAs for functional interrogation (Methods; Supplementary Table S1, Fig. 1a-

c). Using RNA-seq profiling of fractionated RNA we annotated the lncRNA subcellular localization

in the chromatin (35%), nucleus (27%), or cytoplasm (38%) (Fig. 1d). We then designed a

minimum of five non-overlapping antisense oligonucleotides (ASOs) against each lncRNA

(Supplementary Table S2; Fig. 1e) and transfected them individually using an automated cell

culture platform to minimize experimental variability (Fig. 1f).

Fig. 1: Selection of lncRNA targets and the study overview. a, CAGE expression levels (TPM tags per million) and
human dermal fibroblasts (HDF) specificity (Hon, et al., Nature 2017) of lncRNAs in the FANTOM CAT catalog
(n=69,283; grey), lncRNAs expressed in HDFs (n=6,125; blue) and targeted lncRNAs (n=285; red). Dashed vertical
line indicates most lowly expressed lncRNA target (~0.2 TPM). b, Gene conservation levels of lncRNAs in the FANTOM
CAT catalog lncRNAs (grey), expressed in HDF (blue) and targeted lncRNAs (red). Crossbars indicate the median. No
significant difference is observed when comparing targeted and expressed in HDF lncRNAs (Wilcoxon pvalue=0.11).
c, Similar to that in b, but for genomic classes of lncRNAs. Most of the targeted lncRNAs and those expressed in HDF
are expressed from divergent promoters (Hon, et al., Nature 2017). d, Subcellular localization (based on relative
abundances from RNA-seq fractionation data) for targeted lncRNAs. Chromatin-bound (n=98; blue); Nuclear soluble

(n=76; green); Cytoplasmic (n=108; red). Black contours represent all lncRNA expressed in HDF. e, Example of
ZNF213-AS1 loci showing transcript model, CAGE and RNA-seq signal along with targeting ASOs. f, Schematics of
the study.

Knockdown efficiencies had a median value of 34.6%, and for 27% of ASOs, the efficiencies were
higher than 50% (Supplementary Table S2). Overall, ASOs targeting exons or introns were
equally effective, and knockdown efficiencies were independent of the genomic class, expression
level, and subcellular localization of the lncRNA (Supplementary Fig. 1).

Fig.S1: Efficacy of ASOs. Comparison of knockdown efficiencies: a, for exonic and intronic ASOs. b, across different
subcellular localizations of targeted lncRNAs. c, across different genomic classes of targeted lncRNAs. d, across
expression levels of lncRNAs. In all panels, highest knockdown efficiency from the three primer pairs for a given ASO
is shown. Red dot indicates the median value. Knockdown efficiencies with negative values were not considered.

A subset of lncRNAs is associated with cell growth and morphology changes
To evaluate the effect of each lncRNA knockdown on cell growth and morphology, we imaged
ASO-transfected HDF in duplicates every 3 hours for a total of 48 hrs (Supplementary Table S3)
and estimated their growth rate based on cell confluence measurements (Fig. 2a,b). 482 out of
2,021 (~23.8%) ASOs showed significant reduction and 20 (~1%) ASOs showed significant
increase in the growth rate (Supplementary Table S3; *Student's* two-sided t-test FDR < 0.05) as
compared to non-targeting controls (Fig. 2c). Different ASOs targeting the same lncRNA typically
had different effects on growth, possibly due to variable knockdown efficiencies, differences in
targeted lncRNA isoforms, as well as off-target effects. Requiring at least two independent ASOs
to consistently impact cell growth, we identified 10/285 (~3.6%) lncRNAs (Supplementary Table
S3; Z-score < -1.645, FDR < 0.05) involved in maintaining cell growth in HDF. Among lncRNAs
with at least 25% reduced growth rate, we selected *A1BG-AS1*, which was previously implicated
in cell growth (Bai et al. 2018), *CATG00000089639*, *RP11-195F19.9*, and *ZNF213-AS1* (Fig. 2d),

and validated their growth inhibition by measuring the Ki-67 proliferation protein marker upon
knockdown with siRNAs and selected ASOs (Fig. 2e).
In addition to cell growth, we also explored changes in cell morphologies (Fig. 2f). Using a
machine learning-assisted workflow (Methods; Fig. 2g), each cell was segmented and its
morphological features representing various aspects of cell shapes and sizes were quantified
(Carpenter et al. 2006) (Supplementary Table S3). As an example, knockdown of 12/285 lncRNAs
affected the spindle-like morphology of fibroblasts, as indicated by consistent changes in their
observed eccentricity (Z-score < -1.645, FDR < 0.05, at least two ASOs) without reducing the cell
number, suggesting possible cellular transformation towards epithelial-like states. Collectively, we
observed 44/285 lncRNAs (~15.5%) affecting cell growth and/or morphological parameters
(Supplementary Table S3; Fig. 2h).

Fig. 2: Cell growth phenotype assessment. a, Selected example (PTPRG1-AS1) showing the normalized growth rate estimation using NC_A (negative control). b, Correlation of the normalized growth rate for technical duplicates across 2,456 IncuCyte® samples. c, Density distribution of normalized growth rates (technical replicates were averaged) for 2,021 ASO samples targeting lncRNA (red), 270 negative control (NC_A) samples (grey) and 90 mock-transfected cells (lipofectamine only) samples (yellow). Dashed vertical line indicates the 78 (~4%) ASOs with significantly reduced normalized growth rate as compared to other targeting ASOs (ZScore<-1.645; fdr<0.05). d, Growth inhibition for selected lncRNA targets (grey bars indicate matched negative control samples) and the corresponding Incucyte images at 48hrs after transfection. e, Ki67 staining (growth inhibition validation) for selected lncRNA targets after siRNA and ASOs suppression. f, IncuCyte® cell images of a few selected examples of distinct cell

188 morphologies upon the lncRNA knockdown, g, Overview of cell morphology image processing, h, lncRNAs (n=44)
189 significantly (ZScore<-1.645; fdr<0.05) and consistently (at least two ASOs) reducing growth rate (n=10) or/and
190 affecting cell morphologies (n=34).

Molecular phenotyping by CAGE recapitulates cellular phenotypes and highlights functions of lncRNAs

To evaluate transcriptome response upon the knockdowns and its agreement with the observed
cellular phenotypes, we next selected 340 ASOs with high knockdown efficiency (mostly greater
than 50%; median 71.4%) and generated and deep-sequenced 970 CAGE libraries to analyze
154 lncRNAs (Fig. 3a; Supplementary Table S4). Differential gene expression analysis revealed
a median of 15 differentially regulated genes, where 61 (~18%) ASOs yielded zero and 108
(~32%) ASOs yielded 100 or more differentially regulated genes (Supplementary Table S5, Fig.
3b). lncRNAs can recruit a regulatory complex to mediate gene regulation in a site-specific
manner (Long et al. 2017)). To test and identify focal regions of lncRNA-mediated regulation, we
profiled differentially regulated CAGE promoters in genomic-clusters. We applied the Clustering
of genomic REgions Analysis Method (CREAM) (Methods; Supplementary Fig2a; Tonekaboni et
al. 2018) and identified linear clusters were significantly more compact than randomly generated
clusters (Supplementary Fig. 2b; Mann-Whitney $p < 2.2 \times 10^{-16}$) thus supporting biological
relevance of the transcriptome response detected by CAGE molecular phenotyping (Shin et al.
2018). Most identified clusters were distal to the genomic locus of the targeted lncRNA,
suggesting a widespread usage of *trans* regulation mechanisms by lncRNAs (Supplementary Fig.
2c).

Fig. S2: Regulation of gene clusters by lncRNA knockdowns. a, Schematics of finding CORE clusters: differential
transcripts are clustered with CREAM algorithm based on their genomic position. b, Distribution of the CORE clusters
width in the knockdown experiment is narrower than that of clusters generated from randomly generated data. c,

Number of CORE clusters and their relative genomic position to the lncRNA target for each ASO knockdown,
suggesting that lncRNA targets generally affect genes on the same as well as on other chromosomes.

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218 **Fig. 3: CAGE molecular phenotype predicts cellular phenotype.** a, q-PCR knockdown efficiency for 2,021 ASO-
219 transfected samples (targeted lncRNAs). Grey dashed line indicates 50% knockdown efficiency generally required for
220 CAGE selection. Purple dashed line indicates median knockdown efficiency (71.5%) of ASOs selected for CAGE

sequencing. b, Distribution of significantly differentially expressed genes (up-regulated: fdr<0.05, ZScore>1.645, log2FC>0.5 and down- regulated: fdr<0.05, ZScore<-1.645, log2FC<-0.5) across all 340 ASOs. c, Motif Response Activity Analysis (MARA) across 340 ASOs. Scale indicates ZScore of the relative motif activity, with abs(ZScore)>4=4. d, Correlation between normalized growth rate and motif activities across 340 ASOs targeting lncRNAs with highlighted examples Motifs are scaled based on associated TFs expression in HDF (1 to ~600TPM). e, Enriched biological pathways across 340 ASOs. Scale indicates GSEA enrichment value calculated as -log10(pvaule)*sign(NES). f, same as in d, but for selected GSEA pathways. Pathways are scaled based on the number of associated genes.

Next, to assess the specific molecular pathways affected by individual ASOs, we performed: i) transcription factor binding site motif activity response analysis (MARA; FANTOM Consortium et al. 2009) Fig. 3c, ii) gene set enrichment analysis (GSEA; Subramanian et al. 2005; Fig. 3e), and iii) analysis of FANTOM5 co-expression modules (FANTOM Consortium and the RIKEN PMI and CLST (DGT) et al. 2014). Transcription factor motifs that promote cell growth (*e.g.* TFDP1, E2F1,2,3, and EP300) were positively correlated with the measured cell growth rate while transcription factor motifs known to inhibit growth or induce apoptosis (*e.g.* PPARG, SREBPF, and STAT2,4,6) were negatively correlated (Fig. 3d; Supplementary Fig. 3a; Supplementary Table S6). Moreover, correlations between GSEA pathways (Fig. 3f; Supplementary Fig. 3b, Supplementary Table S6) and FANTOM5 co-expression clusters (FANTOM Consortium and the RIKEN PMI and CLST (DGT) et al. 2014); Supplementary Fig. 3c) showed that cell growth and replication related pathways were positively correlated with the measured growth rate, whereas those related to immunity, cell stress and cell death were negatively correlated. Additionally, morphological changes were reflected in the molecular phenotype assessed by CAGE (Supplementary Fig. 3d). Cell radius and axis length were associated with GSEA categories related to actin arrangement and cilia, while cell compactness was negatively correlated with apoptosis. The extensive molecular phenotyping analysis also revealed pathways not explicitly associated with cell growth and cell morphology, such as transcription, translation, metabolism, development and signaling (Fig. 3e).

Fig. S3: Correlations with cell growth and morphologies. A, Correlations of selected transcription factor binding

motifs activity with the normalized growth rate across 340 ASOs targeting lncRNAs. Each hollow circle represents a

single ASO. b, Same as in a, but for selected GSEA pathways. c, Global correlations of FANTOM 5 co-expression

clusters enrichment with growth phenotype. d, Top three significant GSEA pathways positively and negatively

correlated with the normalized growth rate and 13 cell morphologies. Scale indicates FDR-adjusted p value for

Spearman correlation re-scaled through each morphology.

Analyzing 119 lncRNAs targeted by at least two independent ASOs revealed that the

transcriptome response was more concordant for distinct ASOs targeting the same lncRNA than

for ASOs targeting different lncRNAs (Supplementary Fig. 4a). We found 20, 18, and 36 lncRNAs

significantly and consistently changed gene expression, motif activity and biological pathways,
respectively (Methods; Supplementary Table S5; Supplementary Fig. 4b). In total, the molecular
phenotype was affected upon knockdown of 52/119 (43.7%) lncRNAs in human dermal fibroblasts
(Supplementary Fig. 4c).

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Fig. S4: Concordant functional response of lncRNAs. a, Pairwise correlations for CAGE libraries across: 1,000
randomly selected different lncRNAs comparisons, same lncRNAs (different ASOs), same ASOs (duplicates) and 1,000
randomly selected pairwise negative control comparisons. For each data point, the correlations are calculated taking
all genes > 1TPM and dividing their expression by their average expression in all 143 negative controls. b, 52 lncRNA
targets that in at least one functional category (DEG, GSEA, MARA) show a significant consistent response (in at least
2 ASOs) and above generated matched random background (Methods). The significant lncRNA targets in each
category (light bar tip) are summarized in c.

Promoter usage and architecture modulated by lncRNA knockdown

Precise mapping of transcription start sites (TSS) with CAGE previously revealed usage of
alternative promoters and architecture diversities mediated by either TATA or CpG transcriptional
machineries (Carninci et al. 2006; Kim et al. 2005; Haberle et al. 2014; Garieri et al. 2017).
Promoter switching (changes in the *relative* expression between a given pair of promoters of the
same gene) suggest that lncRNAs might direct RNA polymerase II to alternative initiation sites
(Garieri et al. 2017) resulting in different gene isoforms. To test whether we can define regulatory
roles of lncRNAs based on the relative TSS activities in multi-promoter genes, we measured

knowckdown-mediated promoter switching events, which occasionally lead to the change of the
dominant promoter (Fig. 4a-c). For 28 out of 119 lncRNAs, we found 2,699 consistent switching
events (Supplementary Table S7) in 17.8% (1,276 out of 7,161) actively transcribed multiple-
promoter genes (Fig. 4d). The highest number of these events (n=1,159) was found for *CTD-*
2587H24.5 (ENSG00000267577) reported to be a "transcribed enhancer" and a hub in the
chromatin interaction map in K562 cells (Thiel et al. 2018). Knockdown of five other lncRNAs:
AC013394.2, RP11-417E7.1, EMX2OS, ZNF213-AS1, *RAB30-AS1* resulted in 100 or more
promoter switching events (Fig. 4d).

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Fig. 4: Knockdown-mediated promoter usage and architecture. a, Selected example of a promoter switching event
in the *UBC* gene upon the knockdown of *CTD-25877H24.5* lncRNA (left) and of the dominant promoter switching in
DANCR upon its own knockdown (right). b, Number of significant promoter pair switching events (abs(log2FC)>0.5;
FDR < 0.05; abs(ZScore) > 1.645) across all 340 ASOs, with switchings of the dominant promoter highlighted (light
blue). c, 28 out of 119 lncRNA show 2,699 significant and consistent (in at least two ASOs) promoter switching events
affecting 1,276 out of 7,161 expressed multipromoter genes. 196 of these resulted in switching of the dominant
promoter (light blue). d, 2,699 consistent and significant switching events with most prominent lncRNA targets (out of
28 lncRNA targets showing consistent switching) highlighted. e, Up- and down-regulated CAGE promoters with CGI
and TATA-box architectures compared to the background (all non-significantly deregulated promoters). Each data point
represents a single ASO experiment with at least 20 promoters differentially expressed (abs(log2FC) > 0.5; FDR <
0.05; ZScore > 1.645) in a given category.
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Promoter architecture is indicative of gene regulatory machineries with sharp transcription

initiation peaks associated with TATA-box promoters, while broad distributions of transcription

initiation are found in CpG island (CGI) regions. Also, TATA-boxes are often found in promoters

of tissue-specific genes and they are absent in most housekeeping genes (Sandelin et al. 2007).

Thus, the enrichment of TATA-box promoter usage upon knockdown suggest that lncRNAs may

modulate regulatory machineries that govern cell- and pathway-specific promoters and elicit cell-

specific function (Haberle et al. 2014). To assess which lncRNAs influence function in a cell-

specific manner, we quantified the changes in TATA-box and CGI architecture at differentially

regulated promoters and assessed significance for each ASO knockdown (Methods;

Supplementary Table S8). Differentially expressed promoters were more likely to have a TATA-

box both in case of upregulation (two sample Student t-test p-value $< 2.2 \times 10^{-16}$) and

downregulation (two sample Student t-test p-value $= 1.3 \times 10^{-16}$; Fig. 4e). For example, 20-35%

of promoters upregulated upon knockdown of *LINC01615*, *RNASEH1-AS1, RP11-221N13.3*

contained a TATA-box, compared to the 10% of unchanged promoters (Supplementary Fig. 5a,b),

and for *RP11-221N13.3,* additionally, 23-25% of the down-regulated promoters had a TATA-box

architecture (Supplementary Fig. 5b). In contrast, CGI architecture was significantly

underrepresented in upregulated promoters only (stratified sampling p value $< 2.2 \times 10^{-16}$; Fig.

4e). Knockdown of *FGD5-AS1,* for example, showed ~60-70% depletion in two out of three ASOs

as compared to the ~80% of CGI in the background promoters (Supplementary Fig. 5c).

Fig. S5: Promoter architecture usage. a, Selected lncRNAs with significantly up regulated TATA-box promoters. b,

Significantly up and down regulated TATA-box promoters. c, Significantly upregulated CGA promoters. In all panels,

barplots (left) show the percentage of TATA-box or CGI promoters in a given ASO and the numbers represent the

number of differentially expressed promoters. The density plots (right) show the distribution of the stratified random

sampling (10,000 times) for the same number of promoters. The dotted lines indicate observed percentage for each

ASO. The smaller the number of genes randomly sampled, the more discontinuous density distribution.

ZNF213-AS1 is associated with cell growth and migration.

As an example of an lncRNA associated with cell growth and morphology (Fig. 2h) we further

analyzed _ZNF213-AS1_ (_RP11-473M20.14_). This lncRNA is highly conserved in placental

mammals, moderately expressed (~8 CAGE tags per million) in HDF and enriched in the

chromatin. Four distinct ASOs (ASO_01, ASO_02, ASO_05, and ASO_06) strongly suppressed

expression of _ZNF213-AS1_, while expression of the _ZNF213_ sense gene was not significantly

affected in any of the knockdowns. The four ASOs caused varying degrees of cell growth inhibition

(Fig. 5a). ASO_01 and ASO_06 showed a reduction in cell number, as well as an upregulation of

apoptosis, immune and defense pathways in GSEA suggesting cell death. While cell growth

inhibition observed for ASO_02 and ASO_05 was confirmed by Ki-67 marker staining (Fig. 2e),

the molecular phenotype revealed suppression of GSEA pathways related to cell growth, as well

as to cell proliferation, motility, and extracellular structure organization (Fig. 5b), and consistent

in two ASOs downregulation of related motifs, for example, EGR1, EP300, SMAD1..7,9 (Phan et

al. 2004) (Fig. 5c). For data exploration, http://fantom.gsc.riken.jp/zenbu/reports/

As cell motility pathways were affected by the knockdown, we tested whether _ZNF213-AS1_ could

influence cell migration. Using wound closure (scratch) assays with transient cell growth inhibition

(pre-incubation with mitomycin-C and serum starvation) on the transfected cells, we observed a

substantial reduction of wound closure rate (~40% over a 24-hour period) in the _ZNF213-AS1_

depleted HDFs (Fig. 5d). The reduced wound healing rate should thus mainly reflect reduced cell

motility, further confirming affected motility pathways predicted by the molecular phenotype.

Fig. 5: ***ZNF213-AS1* regulates cell growth, migration and proliferation.** a, Normalized growth rate across four distinct ASOs (in duplicates) targeting ZNF213-AS1 as compared to six negative control samples (shown in grey). b, Enrichment of biological pathways associated with growth, proliferation, wound healing, migration and adhesion for ASO_02 and ASO_05. C, Down- and up-regulated binding motifs including examples of those for transcription factors known to modulate growth, migration and proliferation (EGR family, EP300, GTF2I). d, Relative wound closure rate calculated during the 20 hrs post scratching and the representative images of wound closure assay at 20hrs shows 40-45% reduction for the two targeting ASOs (ASO_02 and ASO_05; both in duplicates) as compared to six negative controls samples (shown in grey). e, Knockdown efficiency measured by RT-qPCR after wound closure assay (72 hrs post transfection) showing strong sustained suppression (65-90%) of ZNF213-AS1.

As these results indicated a potential role of *ZNF213-AS1* in cell growth and migration, we used FANTOM CAT Recount 2 atlas (co-submitted to *Genome Research*, GENOME/2019/254656, Imada et al. 2018), which incorporates the TCGA dataset (Collado-Torres et al. 2017), and found relatively higher expression of *ZNF213-AS1* in acute myeloid leukemia (LAML) and in low grade gliomas (LGG) as compared to other cancers (Supplementary Fig. 6a). In LAML, the highest expression levels were associated with mostly undifferentiated states, whereas in LGG, elevated expression levels were found in oligodendrogliomas, astrocytomas, and in IDH1 mutated tumors, suggesting that *ZNF213-AS1* is involved in modulating differentiation and proliferation of tumors

(Supplementary Fig. 6b-e). Further, univariate Cox proportional hazard analysis as well as
Kaplan-Meier curves for LGG were significant and consistent with our findings (HR = 0.61, BH
FDR = 0.0079). The same survival analysis on LAML showed weak association with poor
prognostic outcome but the results were not significant (Supplementary Fig. 6f,g).

Fig. S6: Involvement of *ZNF213-AS1* in cancer. a, Expression of ZNF213-AS1 in TCGA b, Expression of ZNF213-
AS1 for Acute Myeloid Leukemia (LAML) by French-American-British classification. The levels are higher in M0 and
M1 subtypes. These subtypes are considered the most undifferentiated stages of LAML. Since ZNF213-AS1 shows
the lowest expression level in blood cells (GTEx), the higher level of expression could be suggestive of dysregulation
of its expression in undifferentiated stages or involvement in early precursor of blood cells. c, Expression of *ZNF213-*
AS1 in Low Grade Glioma (LGG) by tumor histology classification. Oligodendrocyte differentiation is related to good
prognosis is this cancer type. d, Expression of *ZNF213-AS1* in LGG by tumor IDH (isocitrate dehydrogenase) mutational
status. A total of 511 samples with mutational information were available, 417 of them were classified as "Mutant" and
94 as "wild type". IDH mutation has been repeatedly associated with better prognosis than wild type in gliomas. e,
Expression of *ZNF213-AS1* by patients' vital status. The number of alive patients is 387 from the total of 513 available
in the follow-up. Univariate Cox proportional hazard analysis as well as Kaplan-Meier curves for: f, LGG (HR: 0.61,
FDR: 0.0079) g, LAML (HR: 1.32; FDR: 0.5455)

RP11-398K22.12 (KHDC3L-2) regulates *KCNQ5* in *cis*

Besides primarily observing in *trans* regulation in our knockdown data, several lncRNAs were

previously shown to also regulate the expression in *cis* (Joung et al. 2017). Herein, we

investigated in detail *RP11-398K22.12* (ENSG00000229852) where the knockdowns by two

independent ASOs (ASO_03, ASO_05) successfully reduced the expression of the target lncRNA

(67-82% knockdown efficiency, respectively) and further downregulated its neighboring genes,

KCNQ5 and its divergent partner novel lncRNA *CATG00000088862.1* (Fig. 6a). While the two

genomic loci occupy chromosome 6 and are 650 kb away, Hi-C analysis showed that they are

located within the same topologically associated domain (TAD) and spatially co-localized (Fig.

6b). Moreover, chromatin-enrichment and single molecule RNA-FISH of *RP11-398K22.12* (Fig.

6c) suggested its highly localized *cis*-regulatory role*.*

In FANTOM5 (Hon et al. 2017), expression of *RP11-398K22.12*, *KCNQ5* and

CATG00000088862.1 was enriched in brain and nervous system samples, while GTEx (GTEx

Consortium 2015) showed highly specific expression in brain tissues, particularly in the

cerebellum and the cerebellar hemisphere (Fig. 6d). GTEx data also showed that expression of

RP11-398K22.12 with *KCNQ5* and *CATG00000088862.1* was highly correlated across neuronal

tissues (Fig. 6e,f), with the exception of cerebellum and cerebellar hemisphere, potentially due to

relatively lower levels of *KCNQ5* and *CATG00000088862.1* while levels of *RP11-398K22.12*

remained relatively higher. Additionally, we found an eQTL SNP (rs14526472) overlapping with

RP11-398K22.12 and regulating expression of *KCNQ5* in brain caudate (p = 4.2 x 10^{-6}; normalized

effect size -0.58). All these findings indicate that *RP11-398K22.12* is implicated in the nervous

 system by maintaining the expression of *KCNQ5* and *CATG00000088862.1* in a *cis*-acting

409 manner.

Fig. 6. RP11-398K22.12 mediates the down-regulation of KCNQ5 in cis. a, Changes in expression levels of genes shown by HiC analysis to be in the same topologically associated domain (TAD) as *RP11-398K22.12.* Both *KCNQ5* and *CATG00000088862.1* are down-regulated (pvalue < 0.05) upon the knockdown of *RP11-398K22.12* with two independent ASOs (left) as further confirmed with RT-qPCR (right). b, (top) Representation of the chromatin conformation in the 4Mb region proximal to the TAD containing *RP11-398K22.12,* followed by the locus gene annotation, CAGE, RNA-Seq and ATAC-Seq data for native HDFs. (bottom) Schematic diagram showing Hi-C predicted contacts of *RP11-398K22.12* (blue) and *KCNQ5* (grey) (25Kb resolution, frequency >= 5) in HDF cells. Red line indicates *RP11-398K22.12* and *KCNQ5* contact. c, FISH image for *RP11-398K22.12* suggesting proximal regulation. *TUG1* FISH image (suggesting trans regulation) is included for comparison; (bar = 10 µm). d, GTEx atlas

420   shows relatively high expression of *RP11-398K22.12* in brain samples. e, Correlation of *RP11-398K22.12* and *KCNQ5*

421   in eight distinct brain regions. F, Correlation of *RP11-398K22.12* and *CATG00000088862.1* in eight distinct brain

422   regions.

Discussion

This study systematically annotates lncRNAs through molecular and cellular phenotyping by

selecting 285 lncRNAs from human dermal fibroblasts across a wide spectrum of expression,

conservation levels and subcellular localization enrichments. Importantly, using ASO technology

allowed observed phenotypes to be associated to the lncRNA transcripts, while in contrast

CRISPR-based approaches may synchronically influence the transcription machinery at the site

of the divergent promoter or affect regulatory elements of the targeted DNA site. Knockdown

efficiencies obtained with ASOs were observed to be independent of lncRNA expression levels,

subcellular localization, and of their genomic annotation, allowing us to apply the same

knockdown technology to various classes of lncRNAs.

We investigated the *cis* regulation of nearby divergent promoters, which has been reported as

one of the functional roles of lncRNA (Luo et al. 2016). However, in agreement with previous

studies (Guttman et al. 2011) we did not observe general patterns in the expression response of

divergent promoters. Recent studies suggest that transcription of lncRNA loci that do not overlap

with other transcription unit may influence RNA polymerase II occupancy on neighboring

promoters and gene bodies (Engreitz, Haines, et al. 2016, Cho et al. 2018). Thus, it is plausible

that transcription of targeted lncRNA was maintained, despite suppression of mature or nascent

transcripts using ASOs. This further suggests that the functional responses described in this study

are due to interference of processed transcripts present either in the nucleus, the cytoplasm or

both. While it is arguable that ASOs may interfere with general transcription by targeting the 5'-

end of nascent transcripts and thus releasing RNA polymerase II followed by exonuclease-

mediated decay and transcription termination (aka "torpedo model"; (Proudfoot 2016)), most of

the ASOs were designed across the entire length of the transcript. Since we did not broadly

observe dysregulation in nearby genes, interference of transcription or splicing activity is less

likely to occur.

We observed a reduction in cell growth for ~3.6% of our target lncRNA genes, which is in-line

with previous experiments using CRISPRi-pooled screening, which reported 5.9% (in iPS cells)

of lncRNAs exhibiting a cell growth phenotype (Liu et al. 2017). While these rates are much lower

than for protein-coding genes (Sokolova et al. 2017), recurrent observations of cell growth

(including cell death) phenotypes strongly suggest that a substantial fraction of lncRNAs play an

essential role in cellular physiology and viability.

Several lncRNAs such as *MALAT1*, *NEAT1*, and *FIRRE* have been reported to orchestrate

transcription, RNA processing, and gene expression (Kopp and Mendell 2018), but are not

essential for mouse development or viability. These observations advocate for assays that can

comprehensively profile the molecular changes inside perturbed cells. Therefore, in contrast to

cell-based assays, functional elucidation via molecular phenotyping provides comprehensive

information that cannot be captured by a single phenotypic assay. Herein, differential gene

expression analysis, GSEA (affected pathways) and MARA (changes in transcription factors

binding activity) indicated that 43.7% of lncRNAs exert a regulatory function in HDF. However,

lncRNA targets that did not exhibit a molecular phenotype may be biologically relevant in other

cell types or cell states (Li and Chang 2014; Liu et al. 2017). At the same time, our results showed

that particular lncRNAs expressed broadly in other tissues (e.g., in the human brain) were

functional in HDF (in case of *RP11-398K22.12*), suggesting that lncRNAs may be functionally

relevant across multiple tissues in spite of the cell-type-specific expression of lncRNAs.

An increasing number of studies suggests that lncRNAs regulate gene expression in *trans* via

interaction with chromatin-modifying complexes such as PRC2 (Rinn et al. 2007) or independent

of PRC2 (Portoso et al. 2017) by directly binding to DNA (triplex; (Mondal et al. 2015) or other

RNA binding proteins (Tichon et al. 2016). Analysis of cellular localization by fractionation followed

by RNA-seq and *in situ* hybridization can indicate whether an lncRNA may act *in trans* by

quantifying its abundance in the nuclear soluble fraction as compared to cytoplasm. While most

lncRNAs in nuclear soluble fraction affected pathways associated with chromatin modification,

additional experiments to globally understand their interaction partners will elucidate the

molecular mechanism behind *trans*-acting lncRNAs (Li et al. 2017; Sridhar et al. 2017; companion

paper).

In summary, our study demonstrates that lncRNAs elicit gene regulation largely *in trans* by

regulating the transcriptional machinery for gene activation and repression. These results

highlight the functional importance of lncRNAs regardless of their expression, localization and

conservation levels. Molecular phenotyping is a powerful platform to reveal the functional

relevance of lncRNAs that cannot be observed based on the cellular phenotypes alone. With

additional molecular profiling techniques, such as RNA duplex maps in living cells to decode

common structural motifs (Lu et al. 2016) and Oxford Nanopore Technology (ONT) to annotate

the full-length variant isoforms of lncRNA (Hardwick et al. 2019), structure-to-functional

relationship of lncRNAs may be elucidated further in the future. Data and analysis for this study

is made available at http://fantom.gsc.riken.jp/zenbu/reports/#FANTOM6

Online Methods

Gene Models and lncRNA targets selections

The gene models used in this study were primarily based on the FANTOM CAGE-associated transcriptome (CAT) at permissive level as defined previously (Hon et al. 2017), with additional *de novo* transcript models constructed from human dermal fibroblasts (HDFs) and induced pluripotent stem cells (iPSCs) RNA sequencing data. In brief, CAGE sequencing was performed on the total RNA, and RNA-Seq was performed on ribosomal-RNA depleted RNA, from HDFs and iPSCs (as described elsewhere). CAGE and RNA-Seq reads were mapped onto hg19 using Tophat2 with default parameters. RNA-Seq reads were *de novo* assembled for each cell line using Cufflinks as described previously (Hon et al. 2017) and the transcript models with their 5'ends supported by CAGE reads were retained. LncRNA genes were identified from these retained transcript models as previously described (Hon et al. 2017). The novel lncRNA genes (i.e. loci non-overlapping with FANTOM CAT) were merged with the permissive FANTOM CAT, and the merged assembly were lifted over (Hinrichs et al. 2006) from hg19 to hg38.

From this merged assembly, we selected lncRNA knockdown targets in an unbiased manner to broadly cover various types of lncRNAs. Briefly, we first identified a list of the lncRNA genes expressed in HDF, with RNA-Seq expression at least 0.5 fragments per kilobase per million and CAGE expression at least 1 tag per millions. Then we manually inspected each lncRNA locus in ZENBU genome browser for 1) its independence from neighboring genes on the same strand (if any), 2) support from RNA-Seq (for exons and splicing junctions) and CAGE data (for TSS) of its transcript models and 3) support from histone marks at TSS for transcription initiation (H3K27ac) and along gene body for elongation (H3K36me3), from Roadmap Epigenomics Consortium (Roadmap Epigenomics Consortium et al. 2015). A representative transcript model, which best represents the RNA-Seq signal, was manually chosen from each loci for design of antisense oligonucleotides (ASOs). In total, 285 lncRNA loci were chosen for ASO suppression. Additional controls (NEAT1, protein coding genes Supplementary Table S1) were added including MALAT1 as an experimental control.

ASO design

ASOs were designed as RNase H-recruiting locked nucleic acid (LNA) phosphorothioate gapmers with a central DNA gap flanked by 2-4 LNA nucleotides at the 5' and 3' ends of the ASOs. For each lncRNA target, we used the unspliced transcript sequence from FANTOM CAT as template for designing a minimum of 5 ASOs per lncRNA. A total of 2,055 ASOs targeting 285 lncRNAs were designed for the study (Supplementary Table S1). The gapmer ASOs were between 16 and

19 nucleotides in length (median of 17 nucleotides), had no predicted perfect match off-targets

and 0-1 one-mismatched off-targets, and were designed with a predicted melting temperature

(T_m) in the range of 50-56 °C (Pedersen et al. 2014). The ASOs were synthesized by Exiqon

(~1,500 ASOs) and GeneDesign (~500 ASOs) Inc. and subsequently classified as exonic or

intronic based on overlap with exons in FANTOM CAT gene models.

Cell culture

Human dermal fibroblasts (HDFs) were derived from the dermis of normal human neonatal

foreskin cells (Lonza, catalog number: C2509). The cells were cultured in Dulbecco's Modified

Eagle's medium (high glucose with L-glutamine) supplemented with 10% fetal bovine serum at

37 °C in a 5% CO2 incubator. The passage number of the cells for transfection was maintained

at six or seven.

Automated cell culturing, ASO transfection and cell harvesting

Robotic automation (Hamilton®) was established to provide stable environment and accurate

procedural timing control for cell culturing and transfection. In brief, trypsin-EDTA detachment,

cell number and viability quantification, cell seeding, transfection and cell harvesting were

performed with automation. All transfections were divided into 28 runs at weekly basis. ASO

transfection was performed with duplication. In each run, there were 16 independent transfections

with ASO negative control A (NC_A, Exiqon) and 16 wells transfected with an ASO targeting

MALAT-1 (Exiqon).

The HDF cells were seeded in 12-well plates with 80,000 cells in each well 24 hrs prior to the

transfection. A final concentration of 20 nM ASO and 2 μl lipofectamine RNAiMAX (Thermo Fisher

Scientific) were mixed in 200 μl OptiMEM (Thermo Fisher Scientific). The mixture was incubated

at room temperature for 5 min and added to the cells, which were maintained in 1 ml complete

medium. The cells were harvested 48 hrs post-transfection by adding 200 μl RLT buffer from the

RNeasy 96 Kit (Qiagen) after PBS washing. The harvested lysates were kept at -80 °C.

RNA purification

The harvested lysates were subjected to purification using RNeasy 96 Kit (Qiagen) and epMotion

automated liquid handling systems (Eppendorf) according to the manufacturer's instructions,

except an additional washing step by RPE buffer. The eluted RNA was qualified and quantified

by the Dropsense spectrophotometry platform (Trinean). The RNA was stored at -80 °C.

Real-time quantitative RT-PCR

Real-time quantitative RT-PCR was performed by One Step SYBR PrimeScript™ RT-PCR Kit II

(Takara) using the epMotion automated liquid handling system (Eppendorf). For each sample,

three primer pairs against the specific lncRNA target were used. The expression level was

normalized by GAPDH while the knockdown efficiency was calculated from the fold-change

between each sample and NC_A. The knockdown efficiency of *MALAT-1* was monitored along

each run and across all the runs where all the samples have >90% knockdown. The primers'

sequences are listed in (Supplementary Table S2). In general, the knockdown RNA samples

having >50% knockdown efficiency shown consistently by one of the primer pairs were subjected

to CAGE. Exceptions included insufficient amount of RNA due to cell death and great knockdown

variation between replicates.

ASO transfection for real-time imaging

The HDF cells were transfected manually in 96-well plate to facilitate high-throughput real time

imaging. The cells were seeded 24 hrs before transfection at a density of 5,200 cells per well. A

final concentration of 20 nM ASO and 2 µl lipofectamine RNAiMAX (Thermo Fisher Scientific)

were mixed in 200 µl OptiMEM (Thermo Fisher Scientific). After incubating at room temperature

for 5 min, 18 µl of the transfection mix was added to 90 µl complete medium in each well. The

ASOs were divided in 14 runs and transfected in duplicates. Each plate accommodated 6 wells

of NC_A control, 2 wells of *MALAT1* ASO control and 2 wells of mock-transfection (lipofectamine

alone) control.

Phase-contrast images of transfected cells were captured every 3 hrs for 2 days with 3 fields per

well by the IncuCyte® live-cell imaging system (Essen Bioscience). The confluence in each field

was analyzed by the IncuCyte® software. The mean confluence of each well was taken along the

timeline until the mean confluence of the NC_A control in the same plate reached 90%. The

growth rate in each well was calculated as the slope of a linear regression. A normalized growth

rate of each replicate was calculated as the growth rate divided by the mean growth rate of the 6

NC_A controls from the same plate. Student's t-test was performed between the growth rate of

the duplicated samples and the 6 NC_A controls, assuming equal variance.

Cell fractionation for RNA-sequencing

A previously described method (Conrad and Ørom 2017) was adopted for the isolation of

cytoplasmic, nucleoplasmic and chromatin-associated RNA. Approximately 10 million cells were

used per fractionation experiment. Briefly, trypsinized cells were washed and lysed using cold

lysis buffer containing 0.15% Igepal CA-630, 10 mM Tris pH 7.5, 150 mM NaCl. The lysate was
centrifuged in a sucrose cushion, after which the supernatant was taken as the cytoplasmic
fraction. The nuclear pellet was washed once in buffer containing 20mM HEPES pH 7.5, 50%
glycerol, 75 mM NaCl, 1 mM DTT, 0.5 mM EDTA and suspended again in the same buffer. An
equal volume of nuclear lysis buffer containing 20mM HEPES pH 7.5, 300mM NaCl, 1M Urea,
1% Igepal CA-630, 10mM $MgCl_2$, 1mM DTT, 0.2mM EDTA was added and incubated on ice for
5 min. After centrifugation, the supernatant was considered as the nucleoplasmic fraction and the
pellet as the chromatin fraction. The chromatin pellet was washed once in buffer containing 10
mM HEPES pH 7.5, 10 mM KCl, 10% glycerol, 340 mM sucrose, 4 mM $MgCl_2$, 1 mM DTT and
suspended in the same buffer. RNA from each fraction was isolated using Trizol LS (Invitrogen)
according to manufacturer's instructions. To ensure RNA purity, DNase I treatment followed by
phenol-chloroform extraction was conducted. RNA isolated from each fraction was subjected to
RNA-sequencing.
608
Ki-67 staining upon lncRNA knockdown
For the selected four lncRNA targets showing >25% growth inhibition, we used two siRNAs and
ASOs with independent sequences. The *Silencer® Select siRNAs* were obtained from Invitrogen
and ASOs were from geneDesign (Supplementary Table S9). The HDF cells were transfected
with 20 nM siRNA or ASO in 12-well plates by lipofection. The cells were seeded 24 hours before
transfection at a density of 60,000 cells per well. At 48 hrs post-transfection, cells were washed
by PBS and harvested by trypsin-EDTA. Cells from the two wells with the same transfection were
collected into one tube. After PBS washing, the cells were fixed by adding pre-chilled 70% ethanol
and incubated in -20 °C for at least 2 hours. Ethanol was then removed by centrifugation and the
cells were washed by FACS buffer (2% FBS in PBS, 0.05% NaN3) twice. FITC-conjugated Ki-67
(20Raj1, eBioscience) was applied to cells at a ratio of 8 µl per 150,000 cells. Same concentration
of FITC-conjugated mouse IgG1 kappa antibody (eBioscience) was used as isotypic control. After
1 hr incubation at 4 °C, cells were washed by FACS buffer and subjected to flow cytometric
analysis. Knockdown efficiency by siRNA was determined by real-time quantitative RT-PCR using
the same 3 primer pairs as for ASO knockdown efficiency.
624
Wound closure assay
The HDF cells were transfected by 20nM ASO as described earlier in 12-well plates. The cells
were re-plated at 24 hrs post-transfection into a 96-well ImageLock plate (Essen BioScience) at
a density of 20,000 cells per well. At 24 hrs after seeding, cells form a spatially uniform monolayer

with 95-100% cell confluence. The cells were incubated with 5 µg/mL mitomycin-C for 2 hrs to

inhibit cell division. Then, medium was refreshed and a uniform scratch was created in each well

by the WoundMaker™(Essen BioScience). After changing the medium twice, the cells were

maintained in medium with 0.5% FBS. The condition of mitomycin C and serum concentrations

was tested with HDF and showed complete growth inhibition without severe morphological

change (unpublished data). The closure of the wound was monitored by IncuCyte® live-cell

imaging system (Essen Bioscience) every 2 hrs for 24 hrs. At each time point, the relative wound

density was calculated by the cell confluence within the wound area normalized by the cell

confluence of the non-wound area in the same image. The relative wound closure rate was

calculated as the slope of the linear regression of the relative wound density against time, followed

by normalization with that of NC_A. The RNA was harvested after the assay for real-time

quantitative RT-PCR.

Cell morphology quantification

For each transfection, representative phase-contrast image at a single time point was exported

from the Incucyte time-series, when the confluence of NC_A-transfected cells in each batch was

around 80% (at 30 – 36 hrs post-transfection). These raw images were first transformed to

probability maps of cells by pixel classification using ilastik (v1.3.2)

(https://github.com/ilastik/ilastik.github.io/blob/master/publications.html). Three-pixel categories

including cell, cell boundary, and background were manually labeled in a set of randomly selected

images. The trained model was then applied to all images in the batch mode. The predicted

probability maps of cells (grey scale, 16 bits tiff format) were subsequently used for morphology

quantification in CellProfiler (v3.1.5) (Carpenter et al. 2006). In brief, binarized segmented cell

images were obtained using the module IdentifyPrimaryObjects (thresholding by Global Otsu

method, followed by intensity-based declumping) and the morphology measurements were

performed by the module MeasureObjectSizeShape (http://cellprofiler-

manual.s3.amazonaws.com/CellProfiler-

3.0.0/modules/measurement.html#measureobjectsizeshape). All the values (medians) were

further normalized by those of NC_A from the matching transfection plate, identically to

normalizing the growth rate.

Cap analysis of gene expression (CAGE)

Four micrograms of purified RNA were used to generate libraries according to the nAnT-iCAGE

protocol (Murata et al. 2014). Briefly, random primer with anchor was used for cDNA synthesis,

followed by biotinylation and RNaseI digestion. After cap trapping of the 5′ end complete cDNA

and linker ligation, second-strand was synthesized for dsDNA.Libraries were combined in 8-plex

using different barcodes and were subjected to 50-base single-end sequencing using an Illumina

HiSeq 2500 instrument. Tag were de-multiplexed and mapped to human genome assembly hg38

using TopHat 2.0.12. The average mapping rate was 68.9% with around 10 million mapped

counts obtained on average across all samples.

The samples with mapped counts lower than 500,000 were excluded from further analysis.

Several samples were flagged samples as "questionable" if their mapped counts were less than

1,000,000 or their A260/A230 ratio were less than 1.0. Additionally, we manually flagged as

"questionable" if any exceptional QC metrics were detected after sequencing, which includes

lower amounts of library volumes and possible errors in experiments.

CAGE promoter and gene expression and batch correction

Expression for CAGE promoters was estimated by counting the numbers of mapped tags falling

using 379,953 promoter regions of gene models as described in '*Gene Models and lncRNA*

targets selections'. From there, expression of each of 124,245 genes was estimated by summing

up the expression values of all promoters assigned to a given gene.

Batch correction was performed on the log-transformed 'cpm' values with the prior.count set to

0.25 and normalized for the library sizes using 'removeBatchEffect' function from the 'limma' R

package where the 'batch' was attributed to the CAGE sequencing runs and 'batch2', 'design' and

'covariates' parameters not used.

logCPM <- cpm(dge, log=TRUE, prior.count=0.25, normalized.lib.sizes = T)

removeBatchEffect(logCPM, batch = 'CAGE_lsid')

Differential promoter and gene expression

Differential promoter/gene expression was carried separately for each ASO knockdown (mainly

duplicated CAGE libraries) against matching (the same CAGE sequencing run) negative controls

using DESeq 2.16 with default parameters. For a few ASO knockdowns where CAGE replicates

were in distinct sequencing runs, a generalized linear model (*glm*) with appropriate design was

used. Only promoters/genes with a mean count >=1TPM in either knockdown or negative control

libraries were tested. Promoter/gene was differentially expressed if |log2FC|>0.5; fdr<0.05 and

|ZScore|>1.645; where ZScore was obtained by scaling relative expression change (log2FC) of

each tested promoter/gene was across all experiments.

Motif Activity Response Analysis (MARA)

MARA was performed using batch corrected promoter expression for all the knock-down (KD) and control (both NC_A and NC_B) libraries (970 CAGE libraries). All promoters with expression >=1TPM at least in 70% CAGE libraries (24,014 promoters) were used for the analysis. Transcription factor binding sites (TFBS) for hg38 were predicted as described previously (Arner et al. 2015) using MotEvo (Arnold et al. 2012) for the set of 190 position-weight matrix motifs in SwissRegulon (released on 13 July 2015) (Pachkov et al. 2013) on a multiple alignment of genome assemblies hg38 (human), rheMac3 (macaque), mm10 (mouse), rn6 (rat), bosTau8 (cow), equCab2 (horse), canFam3 (dog), monDom5 (opossum), and galGal4 (chicken). The number of predicted TFBS were counted for each motif in the -300 to +100 base pair from the midpoint of the FANTOM CAT promoters. Next MARA was performed to decompose CAGE expression profiles of the promoters in terms of their associated motifs, yielding the activity profile of all the motifs with at least 150 TFBS associated with the expressed promoters across the HDF KD+controls samples.

Gene Set Enrichment Analysis

Gene set enrichment (GO Biological Process, GO Molecular Function, GO Cellular Component, KEGG, Hallmark and Reactome) analysis was carried separately for each pathway and each ASO knockdown using fgsea (1.8.0) R-package with the following parameters: set.seed(42), minSize=15, maxSize=1000, nperm=100,000, nproc=1. Each pathway was tested based on log2FC preranked genes values from DESEq2 analysis. Pathway was significantly enriched if $|NES|>1$; fdr<0.05 and $|ZScore|>1.645$.; where ZScore was obtained by scaling GSEA significance of enrichment: -log10(pvalue)*sign(NES) of each tested pathway across all knockdowns.

Fantom5 coexpression clusters

Enrichment of Fantom5 coexpression clusters was calculated using fgsea (1.8.0) R-package with the following parameters: set.seed(42), minSize=15, maxSize=5000, nperm=100,000, nproc=1. Coexpression clusters were tested based on log2FC preranked genes values from DESEq2 analysis.

Unsupervised clustering of genomic regions

The unsupervised clustering was performed with CREAM algorithm using all promotes with

log2FC<-1 or log2FC>1 and based on genomic coordinates of dysregulated transcripts.

Cellular and molecular phenotype correlations

Growth and morphology values were correlated for each ASO against 1) each GSEA pathway

using enrichment score, 2) motif activity of each MARA motif (>=1TPM) and 3) each Fantom5

coexpression clusters using enrichment score.

Significant lncRNA targets

The background was estimated 1,000 times for each functional feature (DEG, MARA, GSEA) and

for each lncRNA separately by choosing n random ASOs from n distictint lncRNAs (where n is

the number of all ASOs targeting a given lncRNA of) and, similarly, requiring that at least two of

these ASOs show concordant significance. A target response was considered significant if the

number of DEG or MARA or GSEA was above that estimated from the matching random

background. So for example if a given lncRNA targeted by 3 ASOs has shown 100 DE genes in

any of the two ASOs then it would be significant if 3 randomly chosen ASOs (repeated 1000 times)

had median of less than 100 significant DE genes.

Promoter switching analysis

To identify promoter switching events we used normalized and batch-corrected promoter counts,

considering only genes reaching 4 cpm in at least two of CAGE libraries and promoters reaching

at least 2 cpm in any of CAGE libraries. From this pool, we only considered the ASOs and

promoters that were included into the promoter-level differential expression data. Promoter pairs

passing these expression thresholds corresponded to 7151 genes in total.

For each KD, for expression values of pairs of promoters of the same gene $<p_A, p_B>$, we calculated

their log-transformed ratios $r = \log_2 p_A - \log_2 p_B$, and estimated statistical significance of the

difference $abs(r_{NC} - r_{KD})$ between the data of a target knockdown (KD) induced by a given ASO

and the negative control (NC) libraries. Student's T-test assuming the equal variance in NC and

KD samples was used to evaluated significance. The p-values were FDR-corrected for multiple

testing for the number of tested promoter pairs. Z-score transformation of the log2 fold changes

between r_{KD} and r_{NC} across all ASOs was used to additionally filter by Z-score.

Promoter architecture usage

To identify a CGI overlap and TATA-box presence at ~30 bp upstream of the TSS, CAGE data

were analyzed to identify the precise location of the most frequently used TSS. First, the mapped

reads were imported using CAGEr package (Haberle et al. 2015), and CAGE supported TSSs

(CTSS) were clustered within 20 bp distance, with single CTSS only allowed when > 5 TPM.

These clusters were annotated using FANTOM CAT promoters (CAGE supported promoters).

These were cross referenced with the differential expression gene list per experiment. A negative

control per ASO experiment was chosen randomly to represent core promoter architecture and

each CAGE supported promoter was annotated for a CGI overlapping the promoter and the

presence of a TATA-box (PWM > 80 %) at -35 to -25 bp upstream of the most frequently used

TSS. To avoid an effect of expression levels, we then divided, for each ASO, the CAGE supported

promoters into 5 bins of expression level (low to high) and used stratified sampling (10,000 times)

of the number of promoters in each class of differentially expressed genes separately (up and

down) to assess significance. The background consisted of all promoters per ASO experiment in

the negative control that were not significantly up- or downregulated.

Chromosome conformation capture (Hi-C)

Hi-C libraries were prepared essentially as described previously (Fraser, Ferrai, et al. 2015;

Lieberman-Aiden et al. 2009) with minor changes to improve the DNA yield of Hi-C products

(Fraser, Williamson, et al. 2015). The procedure followed is outlined as a flowchart in

Supplementary Fig. 7a, and each step of the protocol is briefly described below.

Fig. S7: Hi-C protocol overview. a, Flowchart of the Hi-C protocol with related quality control (QC) steps (b-g). b, cDNA recovery quality control. c, PCR and digestion QC. d, DNA shearing QC. e, DNA size-selection QC. f, PCR cycle QC. g, DNA size-selection QC.

Cell cross-linking

Three biological replicate samples of HDFs (~1 X 10^7 cells) grown as described above were fixed

at room temperature in media containing 1% formaldehyde (Sigma, catalog number: F8775) for

10 minutes with gentle rocking every 2 minutes. Cross-linking was stopped by quenching the

formaldehyde with glycine (Sigma, catalog number: G8898) at a final concentration of 125 mM

for 5 minutes at room temperature, followed by 15 minutes on ice. The cells were then scraped

off the plates and pelleted by gentle centrifugation at 400 *g* for 10 minutes. The quenched media

was removed, and the cell pellets were quick-frozen on dry ice before storage at -80°C.

Chromatin digestion

Cell pellets were resuspended into 440 µl of 'Cold Lysis Buffer' containing protease inhibitors (10

mM Tris pH 8.0, 10 mM NaCl, 0.2% Igepal; protease inhibitor cocktail as instructed by the

manufacturer (Sigma, catalog number: P8340). The cells were incubated on ice for 15 minutes

and lysed by twice 20 strokes with a Dounce homogenizer (Pestle B). Cell lysates were

centrifuged at 5,000 rpm for 5 minutes at room temperature in a microcentrifuge. The pellets were

washed twice with 400 µl of 1X NEBuffer^{TM} 3.1, centrifuged 5 minutes at room temperature at

5,000 rpm, and resuspended in 200 µl of 1X NEBuffer^{TM} 3.1. Each sample was divided equally

into 4 tubes before adding 312 µl of 1X NEBuffer^{TM} 3.1 to each tube. 1% SDS (38 µl) was added

to each sample and incubated 10 minutes at 65°C. 10% Triton X-100 (44 µl) was mixed into each

tube before adding 40 µl of NcoI (10 U/µl; 400 Units; NEB, catalog number: R0193L) and

incubating a 37°C overnight.

Biotin labeling and blunt end ligation

Digested fragment ends were labeled with biotin by filling-in NcoI 5' overhangs with Klenow using

biotinylated dCTP. Deoxynucleotides (1.5 µl each of 10 mM dATP, dGTP, dTTP, and 37.5 µl of

0.4 mM biotin-14-dCTP (Life Technologies, catalog number: 19518018)) were added to each

tube, along with 10 µl of Klenow (5 U/µl; 50 Units; NEB, catalog number: M0210S), and incubated

at 37°C for 45 minutes. 10% SDS (86 µl) was mixed into each sample before incubating 30

minutes at 65°C. The samples were transferred to 15 mL conical tubes containing 5.96 mL of

water and 1.66 mL of 'Ligation mix' (10% Triton X-100 (750 µl), 10X Ligation Buffer (750 µl; 0.5

M Tris pH 7.5, 0.1 M MgCl$_2$, 0.1 M DTT), 10 mg/mL BSA (80 µl), 100 mM ATP (80 µl)). T4 DNA

ligase (50 µl) was added to each tube (1 U/µl; 50 Units, Invitrogen, catalog number: 15224-025),

and incubated 4 hours at 16°C.

DNA purification

The DNA was purified by proteinase K digestion followed with phenol/chloroform extraction and

ethanol precipitation. Proteinase K (50 µl of 10 mg/mL) was added to each tube and incubated

overnight at 65°C. Proteinase K was added again on the next day (50 µl of 10 mg/mL) and

incubated for an additional 2 hours. The samples were transferred to 50 mL conical tubes and

extracted with 10 mL of phenol by vortexing 2 minutes and centrifugation at 2,465 *g* for 15 minutes

at room temperature. The samples were extracted again as above but with phenol/chloroform

before precipitating with ethanol. To this end, samples were transferred to 35 mL centrifuge

bottles, and their volumes adjusted to 10 mL with 1X TE before 1 mL of 3 M sodium acetate (pH

5.2) was mixed in and 25 mL of ice-cold ethanol was added. Each sample was gently mixed by

inversion and incubated 1 hour at -80°C. The DNA was pelleted by centrifugation at 23,281 *g*

(13,000 rpm if using SS34 rotor) for 25 minutes at 4°C, washed by resuspending in 10 mL of ice-

cold 70% ethanol, and centrifuged again as above. Pellets were each dissolved into 450 µl of 1X

TE pH 8.0 and extracted twice with 500 µl of phenol-chloroform by vortexing 1 minute and

centrifuging at maximum speed for 5 minutes at room temperature in a microcentrifuge. The

samples were precipitated with 3 M sodium acetate pH 5.2 (40 µl) and absolute ethanol (1 mL),

incubated at -80°C overnight, and centrifuged at maximum speed for 25 minutes at 4°C. The DNA

was washed five times by resuspending in ice-cold 70% ethanol (1 mL), each time centrifuging at

maximum speed for 20 minutes at 4°C. The resulting pellets were each dissolved in 25 µl of 1X

TE pH 8.0 before the 4 samples from original cell pellets were pooled. 10 mg/mL RNAse A (1 µl;

Fermentas, catalog number: EN0531) was added and incubated for 30 minutes at 37°C. The Hi-

C DNA was resolved on 0.8% agarose gel containing ethidium bromide (0.5 µg/mL) for qualitative

assessment of digestion, ligation, and yield based on calibrated molecular weight markers

(Supplementary Fig. 7b).

Quality control

The quality of Hi-C DNA was quantitatively assessed by PCR titration of an expected Hi-C contact

and digestion of the PCR product as previously outlined (Lieberman-Aiden et al. 2009). The Hi-C

DNA was serially diluted in water from 1 to ~0.0005 µL (2-fold; 11 dilutions), and 4 µL aliquots

were PCR-amplified (35 cycles) with primers against a gene desert sequence (GD09 NcoI 5'-

GCAATTAGTGCTATGCCCATGTTTCCTTTGTTCC-3', GD10 NcoI 5'-

CAGTCTTCTACCGCTCTTGTAATGGGGTT-3'). Half of each PCR reaction was resolved on

1.5% agarose gels containing ethidium bromide (0.5 µg/mL) to verify the presence of amplification

products. The remaining products of the first 5 PCR reactions from each titration were pooled,

and divided equally into 4 tubes to digest with either NcoI, NsiI, NcoI/NsiI, or buffer control as

described previously (Lieberman-Aiden et al. 2009). Digestion reactions were incubated 1 hour

at 37°C, and resolved on 2% agarose gels containing ethidium bromide (0.5 μg/mL) to verify

greater digestion efficiency with NsiI. In all 3 HDF biological replicates, over 75% of this PCR

product digested specifically with NsiI (Supplementary Fig. 7c), pointing to efficient Hi-C product

formation in the samples.

Removal of biotin from unligated ends and DNA shearing

Biotin at unligated restriction fragment ends was removed with T4 DNA polymerase as we

described previously (Fraser, Ferrai, et al. 2015). Briefly, 10 samples each of 5 μg DNA from

individual HDF replicates were mixed with 10μl of 10X NEBuffer™ 2.1, 1 μl of 10 mM dCTP, and

5 Units of T4 DNA polymerase (NEB, catalog number: M0203L) in a final volume of 100 μl in

MAXYMum Recovery™ PCR tubes (200 μl; Axygen, part number: PCR-02-L-C). The reactions

were incubated in a thermocycler for 2 hours at 12°C, after which the enzyme was inactivated at

75°C for 20 minutes. The samples were transferred to 1.7 mL tubes (Corning™ Costar™ Low

Binding Plastic Microcentrifuge Tubes, part # C3207), and the DNA was precipitated by mixing

10 μl of 3 M sodium acetate (pH 5.2) in each sample, followed by 275 μl of ice-cold ethanol, and

incubating at -80°C overnight. The DNA was pelleted by centrifugation at 15,000 rpm in a

microcentrifuge at 4°C for 25 minutes, washed twice with 500 μl ice-cold 70% ethanol each time

centrifuging as above, and the resulting pellets were each dissolved in 130 μl of 10 mM Tris pH

8.5. The DNA was sheared to ~350 bp fragments by sonication with a 'Covaris M220 Focused-

utrasonicator' using preset settings (DNA_0300_bp_130μl_Snap_Cap_Micro_TUBE). Samples

were individually transferred to 1.7 mL tubes and shearing to appropriate size range was verified

by resolving 400 ng of each sample onto 1.5% agarose gel containing ethidium bromide (0.5

μg/mL; Supplementary Fig. 7d).

DNA size selection

The sheared DNA was size-selected using AMPure XP Beads (Agencourt XP) with ratios of 0.6X

and 0.85X according to the manufacturer's instructions (Beckman Coulter, catalog number:

A63880). Briefly, 78 μl of beads was added to each 130 μl sample and incubated with rotation for

30 minutes at room temperature. The samples were placed on a DynaMag™ Magnet (Thermo

Fisher Scientific) for 2 minutes, and the supernatant was re-extracted with 32.5 μl of beads as

above to capture the appropriate size fragments. The beads were washed twice with 500 μl of

ice-cold ethanol and air-dried 5 minutes. The DNA was eluted with 60 μl of 10 mM Tris pH 8.5 by

resuspending ten times with a pipette, and the 10 supernatants from each biological replicate

were pooled to a 1.7 mL tube after placing samples on the magnetic stand for 2 minutes. The

sample volumes were then reduced to 300 µl with a SpeedVac concentrator, and 5 µl of each

sample was used to measure DNA concentrations with the Quant-iT™ PicoGreen™ dsDNA Assay

Kit as per the manufacturer's instructions (Thermo Fisher Scientific, catalog number: P11496).

This measure will be used below to calculate the amount of sequencing adaptor required. DNA

recovery was verified by resolving 5 µl of each sample on a 1.5% agarose gel containing ethidium

bromide (0.5 µg/mL; Supplementary Fig. 7e).

Streptavidin pull-down and DNA end repair

The biotinylated Hi-C products were pull-down on Dynabeads MyOne™ Streptavidin C1 Beads

(Thermo Fisher Scientific, catalog number: 65001) to enhance DNA end repair, 3' end

adenylation, and ligation of sequencing adaptors. The beads (60 µl for each Hi-C biological

replicate) were washed with 400 µl 'Tween Wash Buffer' (TWB; 5 mM Tris-HCl pH 8.0, 0.5 mM

EDTA, 1 M NaCl, 0.05% Tween 20) twice for 3 minutes at room temperature with rotation, and

pelleted on a DynaMag™ Magnet for 2 minutes between washes. Beads were next washed once

with 300 µl of '2X Binding Buffer' (2XBB; 10 mM Tris-HCl pH 8.0, 1 mM EDTA, 2 M NaCl), and

resuspended in 300 µl of 2XBB. Each 300 µl-Hi-C library was added to a tube containing beads

and incubated at room temperature for 15 minutes with rotation to bind the DNA. The supernatant

was removed, and the beads were washed twice with 400µl of 1XBB (5 mM Tris-HCl pH 8.0, 0.5

mM EDTA, 1 M NaCl), and once with 100 µl of 1X Ligation Buffer (diluted from 10X, NEB, catalog

number: B0202S). The supernatant was removed and 100 µl of 'DNA End Repair Mix' was added

to each sample (10X Ligation Buffer (10 µl), 10 mM dNTPs (4 µl), T4 DNA Polymerase (5µl of 3

U/µl, NEB), T4 Polynucleotide Kinase (5 µl of 10 U/µl, NEB), Klenow (1 µl of 5 U/µl, NEB), water

(75 µl). The samples were transferred to MAXYMum Recovery™ 200 µl PCR tubes and placed

in a thermocycler for 30 minutes at 20°C. Samples were then transferred to 1.7 mL tubes, and

the beads were washed twice with 200 µl of TWB, and twice with 200 µl Elution Buffer (EB; 10

mM Tris-HCl pH 8.5), each time pelleting beads on a magnetic stand for 2 minutes.

3' end adenylation and sequencing adapter ligation

To prevent concatenation of Hi-C products, DNA fragments were 3' adenylated and ligated to

Illumina Paired-End (PE) sequencing adaptors with 3'-T overhangs. To this end, the supernatant

was first removed from the beads and 50 µl of 3' Adenylation Mix' was added to each tube (10X

NEBuffer™ 2 (5 µl), 10 mM dATP (1 µl), Klenow Fragment (3'→5' exo-; 3 µl of 5 U/µl, NEB, catalog

number: M0212S), water (41 µl). The samples were transferred to MAXYMum Recovery™ 200 µl

PCR tubes and incubated 60 minutes at 37°C in a thermocycler. The samples were transferred

to 1.7 mL tubes, and the beads were washed twice with 200 µl of TWB, and twice with 200 µl EB,

each time pelleting beads on a magnetic stand for 2 minutes.

The amount of Illumina PE Adaptor needed for ligation depends on the amount of DNA in

Hi-C samples. As a rule, 6 pmol of Illumina PE Adaptor (TruSeq™ DNA PCR-Free LT Library

Prep Kit-Set A, Illumina, catalog number: FC-121-3001) was used for every 1 µg of DNA

measured with the Quant-iT™ PicoGreen™ dsDNA Assay Kit after size selection. The EB was

removed from the beads and 45 µl of 'Adaptor Ligation Mix' (5X Invitrogen Ligation buffer (10 µl),

Illumina PE Adaptors, water) was added. Sample were mixed by pipetting, and 5 µl of T4 DNA

Ligase (1 Weiss U/µl, Invitrogen, catalog number: 15224-025) was added to each tube. The

samples were transferred to MAXYMum Recovery™ 200 µl PCR tubes and incubated 60 minutes

at 20°C in a thermocycler. The samples were transferred to 1.7 mL tubes, and the beads were

washed twice for 6 minutes with 400 µl of TWB, and twice with 400 µl EB, each time pelleting

beads on a magnetic stand for 1 minute. Each sample was finally resuspended in 50 µl of EB.

PCR amplification of Hi-C libraries

As a rule, the lowest possible PCR cycle number should be used to amplify enough Hi-C libraries

to reduce amplification biases. To identify the optimal number of PCR cycles, 3 PCR reactions

(25 µl) were prepared for each Hi-C library and amplified through either 6, 8, or 10 cycles.

Individual PCR reactions were composed of 1 µl Hi-C library, 1.25 µl each of 10 µM Illumina PE

1.0 and PE 2.0 primers, 12.5 µl of Phusion High-Fidelity 2X Master Mix (NEB, catalog number:

M0531S), and 9 µl of water. The reactions were conducted in MAXYMum Recovery™ 200 µl PCR

tubes using the following PCR program: 1 cycle of 30 seconds at 98°C, either 6, 8, or 10 cycles

of 10 seconds at 98°C / 30 seconds at 65°C / 30 seconds at 72°C, and a final cycle of 7 minutes

at 72°C. The reactions were resolved on a 2.0% agarose gel containing ethidium bromide (0.5

µg/mL) to select the lowest possible cycle number when a product is detected (Supplementary

Fig. 7f).

Eight PCR cycles was used for large-scale amplification of the libraries for which one set

of 10 PCR reactions for each replicate were prepared as described above except that 1.5 µl of

Hi-C library was used in each reaction. The 10 PCRs from each replicate were pooled and their

combined volume was adjusted to 215 µl with water. The amplified DNA was then purified using

AMPure XP Beads with a ratio of 0.8X by adding 172 µl of washed beads to each pooled reaction

tube and incubating for 10 minutes at room temperature while mixing. The captured DNA was

pelleted on a magnetic stand 2 minutes and washed twice on the beads with 1 mL of ice-cold

70% ethanol. The beads were air-dried for no more than 5 minutes and the DNA was eluted from

the beads by adding 33 µl of 10 mM Tris pH 8.0, 0.1 mM EDTA and pipetting to resuspend the

beads. Sample concentrations were measured using 2 µl of the library and the Quant-iT^{TM}

PicoGreen^{TM} dsDNA Assay Kit as per the manufacturer's instructions (Thermo Fisher Scientific).

DNA recovery and quality was verified by resolving 5 µL of each sample on a 2.5% agarose gel

containing ethidium bromide (0.5 µg/mL; Supplementary Fig. 7g).

Sequencing and processing

The libraries were sequenced on a HiSeq 2500 Illumina platform, and paired-end reads were

mapped to hg38 and processed using HiCUP pipeline ver. 0.5.10 (Wingett et al. 2015). Ditags

were mapped against the human genome assembly hg38. Experimental artifacts such as

circularized, re-ligated, continuous and incorrect size fragments were filtered out. PCR duplicates

were removed from the aligned data. The sequencing and mapping metrics are in (Supplementary

Table S10). Processed mapped ditags from all the three replicates were merged using Samtools

ver. 1.3.1 (Li et al. 2009) for the further analysis. Significant co-localized regions at 10kb resolution

were identified using BioConductor package GOTHiC (Mifsud B., PMID: 28379994). All the

identified co-localized regions with p-value =< 0.01 and fdr =< 0.05 were used for the downstream

analysis. TADs were identified using Arrowhead from Juicer pipeline (Durand et al. 2016). For the

downstream analysis FANTOM CAT promoters were mapped to the colocalized regions and

TADs.

RNA Fluorescence in situ hybridization (FISH)

Oligonucleotide probes against target RNA were designed using the Stellaris Probe Designer

version 4.2 (Biosearch Tech). Probes were flanked on both ends with overhang arms serving as

annealing sites for secondary probes labeled with a fluorescent dye (Chen et al 2015). Overhang

sequences were identical on both ends. Secondary probe sequences have been previously

described (Moffit et al 2016) and were labeled on the 3' end with Atto 647N. All probe sequences

used in this study can be found in (Supplementary Table S11). Two-step hybridization was

performed using a previously-described procedure (Kouno et al. 2019). Briefly, fibroblasts were

seeded onto coverslips overnight and fixed in 4% formaldehyde in PBS for 10 min at room

temperature. Following fixation, coverslips were treated twice with ice-cold 0.1% sodium

borohydride for 5 min at 4 °C. Coverslips were washed three times in PBS, followed by cell

permeabilization in 0.5% Triton X-100 in PBS for 10 min at room temperature. Coverslips were

again washed three times in PBS and treated with 70% formamide in 2X SSC for 10 min at room
temperature. Coverslips were washed twice in ice-cold PBS and once in ice-cold 2X SSC.
Coverslips were either used immediately for hybridization or stored in 70% ethanol for no longer
than a week, in which case were washed in PBS twice and once in 2X SSC at room temperature
prior to hybridization. For hybridization, coverslips were incubated at 37 °C overnight in
hybridization buffer containing 252 mM primary probe inside a humid chamber. Hybridization
buffer contained 10% dextran sulfate, 1ug/ul yeast tRNA, 2mM vanadyl ribonucleoside complex,
0.02% BSA, 10% formamide, 2X SSC. Excess probe was removed by two washes for 30 min at
room temperature in wash buffer containing 30% formamide, 2X SSC, 0.1% Triton X-100,
followed by a rinse in 2X SSC. For second hybridization, coverslips were incubated at 37 °C for
1.5 hrs in hybridization buffer containing 30 nM secondary probe. Excess probe was washed twice
for 20 min at room temperature. Coverslips were stained for 5 min in 2 mg/mL Hoescht in PBS,
washed three times in PBS, and mounted on a glass slide with SlowFade Gold Antifade Mountant
(Invitrogen). Imaging was done on a DeltaVision microscope (GE Healthcare) equipped with a
sCMOS sensor. Image processing was done using FIJI (Chen et al. 2015; Moffitt et al. 2016;
Kouno et al. 2019)
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Author Information

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