Human Endogenous Retrovirus K Rec forms a regulatory loop with MITF that opposes the progression of melanoma to an invasive stage

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Abstract

The HML2 subfamily of HERV-K (henceforth HERV-K) represents the most recently endogenized retrovirus in the human genome. While the products of certain HERV-K genomic copies are expressed in normal tissues, they are upregulated in a number of pathological conditions, including various tumours. It remains unclear whether HERV-K(HML2)-encoded products overexpressed in cancer contribute to disease progression or are merely by-products of tumorigenesis. Here, we focus on the regulatory activities of the Long Terminal Repeats (LTR5_Hs) of HERV-K and on the potential role of the HERV-K-encoded Rec in melanoma. Our regulatory genomics analysis of LTR5_Hs loci indicates that Melanocyte Inducing Transcription Factor (MITF) binds to a canonical E-box motif (CA(C/T)GTG) within these elements in *proliferative* type of melanoma, and that depletion of MITF results in reduced HERV-K expression. In turn, experimentally depleting Rec in a *proliferative* melanoma cell line leads to lower mRNA levels of MITF and its predicted target genes. Furthermore, Rec knockdown leads to an upregulation of epithelial-to-mesenchymal associated genes and to an enhanced invasion phenotype of *proliferative* melanoma cells. Together these results suggest the existence of a regulatory loop between MITF and Rec that

may modulate the transition from *proliferative* to *invasive* stages of melanoma. Because HERV-K(HML2) elements are restricted to hominoid primates, these findings might explain certain species-specific features of melanoma progression and point to some limitations of animal models in melanoma studies.

Introduction

Endogenous retroviruses (ERVs) are remnants of retroviruses that once infected the germline and became vertically inherited as part of the host genome. Sequences derived from various ERVs account for 8% of the human genome [1], reflecting multiple waves of retroviral invasion in the human lineage. In the human genome, HERV-K(HML2) is the most recently endogenized HERV. Although some HERV-K(HML-2) insertions are polymorphic in the human population [2-5], suggesting relatively recent integration events, no fully-active provirus has been identified. Due to recombination events and other post-insertional mutations, the vast majority of HERV-K(HML2) members (~950-1200) are either solitary long terminal repeats (solo LTRs) or variously truncated copies [6, 7]. A subset of proviral copies (~90) are still capable of expressing retroviral proteins (e.g., Gag, Pro-Protease, Pol-Polymerase, Env-Envelope), and can apparently produce viral particles [8-10]. HERV-K(HML2) proviruses have been further classified as type 1 (~26%) or type 2 (~74%), depending on the presence or absence of a 292-bp deletion overlapping the Pol-Env junction [11]. Np9 and Rec are alternatively spliced products of Env [8, 12]. Type 1 proviruses may only encode Np9 [12], whereas type 2 proviruses may express the RNA isoforms encoding Env, Rec, and Np9 [13-15]. Based on the phylogeny of their LTR sequences, HERV-K(HML-2) can be further classified into LTR5_Hs, LTR5_A/LTR5_B subfamilies [16, 17]. Elements bearing LTR5_Hs represent the youngest integration events in the human genome; with 615 copies (45 proviruses and 570 solo LTRs), 90% of them being exclusive to humans, and absent in other great apes [16]. Type 1 and 2 proviruses are about evenly split among these elements, with 20 and 17 copies respectively [9].

Rec is a small RNA-binding protein considered to be a functional homolog of the HIV
Rev accessory protein [13]. Rec contains a nuclear localization signal (NLS), which in
addition to facilitating nuclear transport of the Rec protein, recognizes the Rec-
responsive element (RcRE) within the viral RNAs [14, 18]. Binding of Rec to RcRE
stabilizes incompletely spliced/unspliced viral transcripts and enhance their nuclear
export [8, 19, 20] via the CRM1 export pathway [12, 13, 21]. Moreover, Rec interacts
with Staufen-1 to further facilitate unspliced viral RNA transport [22]. Furthermore, Rec
can directly modulate cell signalling via binding to the promyelocytic leukemia zinc
finger (PLZF) protein, a transcriptional repressor of the c-MYC proto-oncogene [23].
Rec also binds the TZFP (testicular zinc-finger protein) and the hSGT (human small
glutamine-rich tetratricopeptide repeat protein), which are transcriptional repressors of
the androgen receptor [22, 24]. While it is evident that Rec must have first evolved to
facilitate HERV-K replication, there is speculation that the protein may have been co-
opted for physiological function(s), such as spermatogenesis [21, 25] and defence
against viral infections in the early human embryo [26]. Although Np9 is expressed in
various types of cancer [12, 27] and healthy tissues [28], unlike Rec, it has no identified
function promoting the replication of HERV-K [29].
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It has been reported that at least 18 HERV-K genomic loci can be transcribed in healthy
human tissues and express potentially-coding Rec or Np9 mRNAs [28]. However, the
landscape of transcribed HERV-K loci differs considerably between cell/tissue types
[28]. Rec RNA/protein expression has been observed in human embryonic tissues,
placenta and retina [15, 30]. Early embryonic cells appear favourable to HERV-K
expression as proviral mRNAs and proteins are relatively abundant in both germinal-
and pluripotent stem cells [26, 31]. While HERV-K expression has been observed at
low levels in various somatic healthy tissues, it can be further induced by environmental
stress, such as ultraviolet irradiation (e.g., UVB and UVC), starvation, or viral infection
[32-36]. Furthermore, many studies have reported robust mRNA upregulation of
several genomic loci of HERV-K in various disease states, including certain
autoimmune diseases and several cancers, especially lung, breast cancers, germ cell
tumours and melanoma [33, 37, 38] and reviewed in [39, 40]. Moreover, transcriptional
activation and translation of various HERV-K products [37, 41] and even virus-like
particles have been observed in specific cancer cells such as teratoma, embryonic
carcinoma and melanomas [8, 10, 42, 43].

Whether and how the overexpression of HERV-K products may contribute to the etiology or progression of these diseases has been the subject of many studies, debates, and speculations. Several mechanisms have been proposed by which HERV-K might contribute to disease. Around 50% of the HERV-K LTRs (including solo LTRs) retained some of their promoter/enhancer activity [44], and are differentially expressed in healthy and cancer tissues. The activated LTRs could act as alternative promoters and de-regulate tumour suppressor genes or proto-oncogenes (reviewed in [7, 45]). Furthermore, the Env protein of HERV-K, via its fusogenic property, is capable of inducing cell-cell fusion and could, therefore, contribute to tumour invasiveness [27, 33, 46]. Env could also have oncogenic properties through direct interference with cell signalling pathways (e.g., RSS/MEK/ERK) [47, 48]. Indeed, the overexpression of the HERV-K Env protein has been reported to induce an epithelial to mesenchymal transition (EMT)-like process [47, 49], a crucial event in oncogenesis leading to the more malignant phenotype. In addition, Env was suggested to promote tumorigenesis via modulation of the immune response [50-52]. Confusingly, HERV-K may have either a positive or negative effect on the immune system [53, 54]. Similar to Env, the expression of Rec and Np9 is significantly increased in several pathological conditions [55], including germ cell tumours and melanoma [12, 21, 37, 42, 43, 56-60], but again it remains unclear whether and how these factors could contribute to tumorigenesis and other disease states.

To begin answering this question, we investigate the regulatory genomic aspects of LTR5_Hs sequences, as well as the oncogenic properties of HERV-K Rec (Rec thereafter) in melanoma. We used *in vitro* models as well as high throughput data mining, including single-cell transcriptome analyses of patient samples, to evaluate the impact of Rec expression in the spatiotemporal progression of melanoma. We found that Rec marks the *proliferative* state of melanoma, and similarly to Env [47] modulates the EMT-like process of cell transformation. However, in contrast to Env [47], Rec appears to inhibit the transition from proliferative to invasive state and as such may represent a protective factor in melanoma.

2. Materials and Methods

 2.1

Methods:

Generating the KD-Rec cell lines

All 3 KD targeting the ERVK6(HML2.HOM) locus. KD1 targets 3'UTR, KD2 3'UTR, and

KD3 the gene body of Rec (**Figure S2**). **KD1** (as) 5'-ATCCATTCAACTCTGAGTGGA-

3', chr7:4,572,718-4,593,717; **KD2** (as) 5'-TAAGGCTGACTGTAGATGTAC-

3'chr7:4,580,334-4,601,333, chr7:4,599,327-4,599,347; **KD3** (as) 5'-

CAACGGTGCTCGATTGCGATG-3' chr7:4,585,379-4,585,399, chr7:4,593,883-

4,593,903. Note that due to the high similarity of the HML2 sequences, other loci are

also likely to be targeted. The control (scrambled) KD was designed to target GFP

cDNA, GCGAAGTACGAATAGTTATCA from [61]. To generate stable KD clones, the

targeting sequences were cloned into pT2-MP71-(KD-Scr/1/2/3)-EGFP construct [62],

and co-transfected with SB100X transposase [63] into A375 and SKMel-28 cells. The

stable lines, which should also express a green fluorescent protein (GFP) reporter,

were selected using two-rounds of FACS sorting for the GFP-reporter signal.

Cell Culture

The melanoma cell line, A375 (ATCC CRL-1619) was cultivated in RPMI1640 (Thermo

Fisher Scientific) supplemented with 10% fetal bovine serum (FBS; PAA Laboratories,

Pasching, Austria), 80 U/mL penicillin and streptomycin (Lonza, Basel, Switzerland) at

37°C in a humidified atmosphere of 95% air and 5% CO_2. The melanoma cell line,

SKMel-28 (ATCC HTB-72) was maintained in Dulbecco's modified Eagle's medium

(DMEM, Life Technologies, England) containing GlutaMAX™-I supplement (Gibco,

catalogue number: 31966-021), with 10% FBS and penicillin and streptomycin in the

same condition as above.

Transfection of melanoma cells and cell sorting

To stably knock down (KD) Rec, A375 and SKMel-28 melanoma cell lines were

transfected with pT2-MP71-(KD1-Rec)-EGFP, pT2-MP71-(KD2-Rec)-EGFP, pT2-

MP71-(KD3-Rec)-EGFP plasmids in the presence of pT2-CAGGS-SB100X, using the

Neon transfection system (Life technologies). The electroporation setting was 1350 V

(pulse voltage), 20ms (pulse width), 2 pulses. $2×10^5$ cells in 11 µl resuspension buffer

R were combined with 2 µl of purified plasmid mixture (50 ng transposase and 500 ng

transposon). Transfected cells were transferred to 6-well tissue culture plates

containing 2 ml culture medium. EGFP+ cells were sorted at weeks 1 and 3 post-

transfection by FACSAria II cell sorter (Becton).

Quantitative Real-Time PCR and semi-quantitative PCR

Total RNA was extracted using the Direct-zol RNA MiniPrep kit (Zymo Research)

according to the protocol of the manufacturer. Reverse transcription was performed

from 1 µg total RNA with the High Capacity RNA-to-cDNA kit (Applied Biosystems).

The isolated RNA would be treated with additional DNaseI (Invitrogen), to get rid of

potential DNA contamination. Quantitative real-time PCR (qRT-PCR) was carried out

on ABI7900 with PowerUp SYBR Green Master Mix kit (Applied Biosystems),

according to the recommendations of the manufacturer. For quantification of mRNA

expression levels, real-time PCR was run in triplicates for each cDNA sample, using

GAPDH and 18s RNA as the internal controls for SKMel-28 and A375 cells,

respectively. For primers and annealing temperatures, see **Table 1**. Data were

analysed using the comparative CT ($2-\Delta\Delta CT$) method, which describes relative gene

expression. For RNA-seq validation, the qPCR control was transfected with pT2-CAG-

GFP instead of the KD constructs.

Table 1. The primers/oligos used in this study.

Name	Application	Sequence (5' – 3')	Tm(°C)
HML-2-Env	Real-Time qPCR	F: GCTGCCCTGCCAAACCTGAG R: CCTGAGTGACATCCCGCTTACC	60
HML-2-Gag	Real-Time qPCR	F: AGCAGGTCAGGTGCCTGTAACATT R: TGGTGCCGTAGGATTAAGTCTCCT	60
HML-2-Np9	Real-Time qPCR	F: AGATGTCTGCAGGTGTACCCA R: CTCTTGCTTTTCCCCACATTTC	60
HML-2-Rec	Real-Time qPCR	F: ATCGAGCACCGTTGACTCACAAGA R: GGTACACCTGCAGACACCATTGAT	60
MITF(M)	Real-Time qPCR	F:ATGCTGGAAATGCTAGAATATAATCACT R: GAATGTGTGTTCATGCCTGG	60
SPANXB1/B2	Real-Time qPCR	F: AGGCCAATGAGGCCAACAAGAC R: TCCTCCTGTAGCGAACCACTAG	60

18S RNA	Real-Time qPCR	F: GTAACCCGTTGAACCCCATT R: CCATCCAATCGGTAGTAGCG	60
GAPDH	Real-Time qPCR	F: CAATGACCCCTTCATTGACCTC R: AGCATCGCCCCACTTGATT	60

RNA sequencing and data analysis

The concentration of RNA was measured on NanoDrop Spectrophotometer ND-1000,

and the quality of RNA was analysed using Agilent RNA 6000 Nano Kit on Agilent 2100

Bioanalyzer machine. The RNA sequencing library was prepared from 550 ng of RNA,

using Illumina TruSeq Stranded mRNA LT Set A kit (Cat. no. RS-122-2101), according

to TruSeq Stranded mRNA Sample Prep LS Protocol. Sequencing was performed on

an Illumina HiSeq 2000 platform as 100 bp first strand-specific paired-end reads.

Sample-specific barcoded sequencing reads were de-multiplexed from multiplexed

flow cells and by using CASAVA. 1.8.2 BCL files were converted to FASTQ format

files. The quality of the raw sequence reads was determined by using the FastQC.

Reads with a quality score below 30 were removed. We removed the highly variable

two nt from the ends of the remaining sequencing reads and mapped them over the

reference genome (Human hg19/GRCh37) and transcriptome model (hg19.refseq.gtf).

hg19/GRCh37 and hg19.refseq.gtf that were downloaded from USCS tables

(http://hgdownload.cse.ucsc.edu/goldenPath/hg19/bigZips/). For mapping the reads,

we used our defined settings (i.e. *–alignIntronMin 20 –alignIntronMax 1000000 –*

chimSegmentMin 15 –chimJunctionOverhangMin 15 –outFilterMultimapNmax 20) to

STAR splice mapper. This approach generated STAR genome/transcriptome indices

of the reads and provided their respective RefSeq gtf (genes and rmsk) annotations.

Final read alignments having more than two mismatches per 100 bp were discarded.

We obtained uniquely mapped read counts over individual gene bodies using

featureCounts function from the subread package, at gene and transcript level using

RefSeq annotations. Scalable gene expression levels were calculated as transcript per

million (TPM) counts over the entire gene (defined as any transcript located between

transcription start and end sites). In the datasets generated in different batches, we

normalized the batch effect using the 'RUVSeq' package from R. To obtain differentially

expressed genes (DEGs). Firstly, we removed the extremely low expressing genes

(genes with less than five mapped reads) to avoid the unnecessary variation in

transcriptomes. We then used *DESeq2*, a R package, to determine the differentially
expressed genes from replicated data. Data were normalized using the default
functions of DESeq2. The package "*DESeq2*" provides methods to test for differential
expression by use of generalized linear models; the estimates of dispersion and
logarithmic fold changes incorporate data-driven prior distributions. Genes with
adjusted p-value less than 5% (according to the FDR method from Benjamini-
Hochberg), Log2 Foldchange |1| were assigned as differentially expressed. This
strategy provided both quantification and statistical inference of systematic changes
between conditions (with at least three replicates). The following samples were RNA
sequenced: KD-Scr_A375, KD1-Rec_A375, KD2-Rec_A375, KD3-Rec_A375, KD-
Scr_SKMel28, KD1-Rec_SKMel28, KD2-Rec_SKMel28, KD3-Rec_SKMel28 in two
batches. Canonical pathways and biological function of the differentially expressed
genes (DEGs) were further subjected to KEGG and Gene Set Enrichment Analyses
(GSEA) using the webgestalt tool.
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To estimate expression levels for repetitive elements, we calculated the CPM (counts
normalized per million of total reads mappable on the human genome) for each Class
II TE locus given in the repeat-masked annotations, downloaded from the UCSC data
portal. We had removed those repetitive element's locus that overlays with the genic
exons/UTRs or in the vicinity of 1 Kb from the TSS of annotated genes before mapping
the sequencing reads over the individual locus. To calculate the family-wise expression
of repetitive elements, we considered multi mapping reads only if they were mapping
exclusively within a repetitive family. Then counted one alignment per read to calculate
the CPM. The expression level of repeat families is calculated as Log2(CPM+1) prior
to comparison. The locus-specific information was only calculated for those HML2-
HERV-K loci that encodes for Rec transcripts. For this, we used "Salmon" that
implements the expectation-maximization algorithm to re-distribute and assign the
multi-mapping reads to the provided coordinates, based on evidence from uniquely
mapped reads.
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Single-cell RNA-seq data processing
We used the single-cell count matrix from GSE72056. We calculated the activity of
genes in every cell at TPM expression levels. We considered samples expressing over
1000 genes with expression levels exceeding the defined threshold (Log2 TPM > 1).

We used Seurat_3.1.1, a package from R to normalize the datasets at the logarithmic

scale using "scale.factor = 10000". After normalization, we calculated scaled

expression (z-scores for each gene) for downstream dimension reduction. We used

the original annotations of datasets to classify the Malignant, Non-malignant and

Heterogenous cell-types. To define cell population clusters, we employed the

FindClusters function of "Seurat" using "PCA" as a reduction method. The specific

markers for each cluster identified by "Seurat" were determined by the

"FindAllMarkers" function, using "roc" as a test for significance. This provided us two

lists of gene sets. 1) Malignant vs. Non-Malignant differentially expressed genes and

2) genes differentially expressed between MITF-High and MITF-Low tumours, which

we applied for comparison with Rec-KD RNA-seq DEGs. Feature plots, violin plots and

heatmaps were constructed using default functions, except for the colour scale that

was set manually. The annotated cells were re-clustered using the methodologies

described above and visualized on the UMAP coordinates.

ATAC-seq and ChIP-seq data analyses

ATAC-seq and ChIP-seq raw datasets in sra format were downloaded from the listed

studies and converted to fastq format using sratools function fastq-dump –split-3.

Fastq reads were mapped against the hg19 reference genome with the modified

bowtie2 parameters: --very-sensitive-local. The modified settings are the following: -D

(number of matches) was set to 30, N (number of mismatches in seed) was set to 1,

and -R (number of re-seed) was set to 5. All unmapped reads with MAPQ < 10 and

PCR duplicates were removed using Picard and samtools. MACS2 called all the

ATAC-seq peaks with the parameters –nomodel -q 0.01 -B. Blacklisted regions were

excluded from called peaks

(https://www.encodeproject.org/annotations/ENCSR636HFF/). To generate a set of

unique peaks, we merged ATAC-seq peaks using the mergeBed function from

bedtools, where the distance between peaks was less than 50 base pairs. We then

intersected these peak sets with repeat elements from hg19 repeat-masked

coordinates using bedtools intersectBed with a 50% overlap. To calculate the

enrichment over the given repeat elements, we first extended 5 Kb upstream and 5 Kb

downstream coordinates from the left boundary (TSS) of respective elements in a

strand-specific manner. These 10 Kb windows were further divided into 100 bps bins

and tags (bedGraph output from MACS2) were counted in each bin. Tag counts in

each bin were normalized by the total number of tags per million in given samples and
presented as CPM per 100 bps. We averaged CPM values for each bin between
replicates before plotting the figures. To find the TF binding motifs, 25 bps sequences
were extended from either side of MITF ChIP-seq peak summits. The extended
sequences were analysed by RSAT tool (http://rsat.sb-roscoff.fr/). The TF binding
motifs were calculated from JASPAR libraries of human TF motifs.
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Cell invasion assay
The cell invasion assays were performed using transwell chambers (8 μm pore size;
Corning Costar) according to the vendor's instructions. Briefly, the insert of the wells
was first coated with 50 μl BD matrigel. A375 cells were resuspended with RPMI1640
medium containing 0.5% FCS to reach a concentration of 5×10^5 cells/ml. Then 100 μl
cells from each group (stable KD-REC clones, pT2-CAG-GFP control) were seeded
into the upper chamber of the insert, with adding DMEM containing 10% FCS to the
lower chamber. After incubation at 37 °C in a humidified atmosphere of 5% CO_2 for 24
h, non-invaded cells were removed with gentle swabbing. In contrast, the invaded ones
were stained, and then images were taken with microscopy. Cell numbers were
calculated with ImageJ. The control was transfected with pT2-CAG-GFP instead of the
KD constructs.
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The accession of datasets used in this study:
Encode TF ChiP-seq datasets:
ftp://hgdownload.cse.ucsc.edu/goldenPath/hg19/encodeDCC/wgEncodeSydhTfbs
GSE60663: H3K27Ac Primary Melanoma
GSE60663: H3K4Me3 Primary Melanoma
GSE60663: MITF ChIP-seq Primary Melanoma
GSE60664: RNA-seq Primary Melanoma
GSE46817: RNA-seq of 7 distinct melanoma lines
GSE46805: RNA-seq Melanocyte BRAF overexpression
GSE50681: MITF ChIP-seq Melanoma and Melanocyte post-treatment
GSE82330: ATAC-seq Melanoma A375
GSE72056: Patient's Single-cell RNA-seq data
GSE61966: MITF knock-down RNA-seq in 501 Melanoma and Melanocytes
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3. Results

3.1. LTR5_Hs loci form active chromatin in cancer and pluripotent cell lines

Cis-regulatory elements driving the transcription of the youngest subfamily of HERV-K are embedded within the human-specific LTR5 (LTR5_Hs) [64, 65]. Various transcription factors (TFs) have been reported to bind LTR5_Hs, including the pluripotency factors such as OCT4 [66] or the melanocyte inducing TF (MITF) [65, 67]. To examine more systematically the regulation of HERV-K transcription in different cell types, we mined publicly available data produced by the ENCODE consortium [68]. Briefly, we mapped the peaks from 142 ChIP-seq TF datasets from 7 cell lines (ENCODE) over 615 genomic sequences annotated as LTR5_Hs in the human genome (see **Methods**). Our integrative analysis of various TF occupancy revealed that multiple TFs bound a group of LTR5_Hs loci in a cell-type specific fashion. Notably, in H1 embryonic stem cells (H1ESCs) and leukemia K562 cells, LTR5_Hs elements tend to be bound by more TFs than in the other cell lines examined (**Figure 1**A). In the leukemia cell line K562, LTR5_Hs elements are enriched for binding of multiple TFs that control cell cycle and proliferation (e.g., cMYC, MAX, JUND, PU1 and P300, see **Figure 1**A), which might reflect the transformed phenotype of these cells. Curiously, NANOG binding appeared particularly prominent when H1 cells were cultured in naïve ("3iL") conditions [69] (**Figure 1**B).

3.2. MITF-regulated LTR5_Hs/HERV-K expression is a hallmark of the 'proliferative' type of melanoma

It has been previously reported that in melanoma, the melanocyte-specific isoform MITF(M) is expressed and promote the expression of HERV-K transcripts likely through direct binding to LTR5_Hs [67]. To further substantiate that MITF drives HERV-K expression in melanoma, we examined the expression levels of various transposable elements (TEs) in MITF-depleted melanocytes and melanoma cells using previously published RNA-seq datasets [70]. Knocking down MITF affected the expression of a small subsets of TE families that were differentially expressed in both melanocytes and melanoma, but there was only 5% overlap in differential TE expression between the two cell types (**Figure 1**C-D). Notably, KD MITF was

associated with a significant decrease in HERV-K expression (adjusted p-value < 0.01,

Benjamini-Hochberg (BH) correction) in melanoma, but not in melanocytes (**Figure**

1C), suggesting that MITF may regulate HERV-K expression specifically in

melanomas.

MITF(M) is a critical transcription factor associated with melanoma progression [71,

72], and is a sensitive marker of melanoma invasiveness: it is highly expressed (e.g.,

MITF-High) in the melanocytes and the *proliferative* state of melanoma, but lowly (e.g.,

MITF-Low) expressed in the *invasive* state [73]. To investigate how tightly HERV-K is

controlled in melanoma by MITF, we mined ChIP-seq and RNA-seq data from

melanocytes, various melanoma cell lines, and primary melanoma cultures (**Methods**).

Characterizing primary melanoma cultures according to their MITF levels (**Figure 1**E)

revealed an inverse correlation between LTR5_Hs/HERV-K RNA levels and

invasiveness (**Figure 2**A). To see if additional TE families respond to melanoma

proliferative/invasive status, we performed a more systematic analysis including other

TE families. This analysis revealed that the expression of certain ERVs, including

LTR5_Hs/HERV-K, LTR2C and LTR13, was correlated with that of genes

characterizing the *proliferative* state of melanoma (**Figure S1**A). When compared to

the overall expression of other TE families, LTR5_Hs/HERV-K was the most

abundantly expressed TE family and showed the strongest positive correlation with

MITF levels (rho = 0.43, corrected p-value < 0.2e-9) in the melanoma primary cultures

(n=10) (**Figure S1**B). These data suggested that similarly to MITF, LTR5_Hs/HERV-K

expression correlated negatively with melanoma invasiveness. Next, we characterized

the regulatory activation of LTR5_Hs in the *proliferative* type of melanoma cells by

layering ChIP-seq occupancy signals of MITF and H3K4Me3, H3K27Ac (**Figure 2**C

and **S1**C-D). Relative to the invasive state, the proliferative state showed enrichment

of LTR5_Hs for these "active" histone marks, coupled with MITF binding (**Figure 2**C

and **S1**C-D). We found that MITF binding at LTR5_Hs loci coincided significantly (p-

value = 0.026, chi-square test of Observed vs. Expected ratio) with the presence of the

canonical E-box motif (CA(C/T)GTG), which is recognized by MITF as well as c-MYC

and MAX TFs [74] (**Figure 2**C). Overall, our analysis suggests that the upregulation of

HERV-K expression in *proliferative* melanoma is likely driven at least in part by direct

binding of MITF over the LTR5_Hs through the E-box motif.

3.2. Inhibition of BRAF^{V600E} mutant leads to the MITF binding over LTR5_Hs/HERV-K in melanoma

We also investigated the transcriptional regulation of HERV-K in the background of the BRAF^{V600E} mutation, which characterizes an *invasive* type of melanoma [75]. PLX4032, a BRAF^{V600E} inhibitor commonly used in the clinic for the treatment of melanoma, promotes the expression of MITF by several folds and causes a reversion of melanoma from *invasive* to *proliferative* state [75]. To examine how PLX4032 treatment may affect HERV-K regulation, we reanalysed publicly available MITF ChIP-seq datasets of primary melanocytes and a BRAF^{V600E} mutant melanoma cell line (COLO829) with and without PLX4032 treatment (GSE50681). We observed that PLX4032 treatment results in a substantial increase of MITF ChIP-seq peaks over LTR5_Hs elements (123 out of 615 loci were bound under PLX4032 treatment but only 19 were bound in the control cells) (**Figure 2**D), suggesting that many LTR5_Hs loci are bound by MITF when it is highly expressed. Among all TE families in the human genome, the gain of MITF binding under PLX4032 treatment was unique to the LTR5_Hs family (p-value < 0.01, chi-square test of observed vs. expected ratio) (**Figure 2**E). While MITF is also expressed and binds to a subset of LTR5_Hs elements in healthy pigmented melanocytes (n=33), there was no elevated MITF ChIP-seq signal over LTR5_Hs elements upon PLX4032 treatment of those cells (**Figure 2**D). Thus, the *proliferative* state of melanoma is associated with robust and widespread binding of LTR5_Hs by MITF.

Next, we seek to gain insight into the transcriptional activity of individual HERV-K loci in different melanoma states. Because available RNA-seq datasets were all using short reads, which precludes locus-specific analysis of recent TEs like HERV-K, we turned to chromatin ChIP-seq data (which is more mappable) as a proxy to predict transcriptionally active HERV-K loci [76]. Namely, we used published dataset for H3K27ac and MITF ChiP-seq (GSE60663) for low passage melanoma cells and also ATAC-seq and H3K27Ac data for the A375 melanoma cell line (GSE82334). Using this approach, we were able to identify twelve loci (nine HERV-K proviruses and 3 solo LTR5_Hs) bound by MITF and/or with active chromatin marks (H3K27ac, ATAC peak) in MITF-High primary cells (but not in MITF-Low cells) (**Table S2**). For example, **Figure 3**A presents an illustration of a region on chromosome 7 that includes the ERVK6 locus, which is structured as two nearly-identical tandemly arranged proviruses [77], mirrored around a centrally located third LTR5_Hs, encoding for apparently intact

ORFs [28]. The flanking LTR5_Hs of the tandem proviruses showed higher H3K27Ac
signal in both *proliferative* MITF-High and A375 cells relative to MITF-Low cells, and
also significant ATAC-seq and H3K27ac signals in A375 (**Figure 3**A), indicating that
these loci are preferentially active in the *proliferative* type of melanoma. Moreover,
genome-wide analyses of ATAC-seq and active histone marks (H3K27ac and
H3K4Me1) in A375 cells further support the notion that multiple LTR5_Hs elements
(>50) may function as promoters or enhancers in this melanoma cell line (**Figure 3**B-
C). Collectively, these regulatory genomics data suggest that MITF binding contributes
to the activation of multiple LTR5_Hs loci copies (both solo and proviral) in the
proliferative type of melanoma.
391
392   *3.3. Depleting HERV-K-Rec may induce an EMT-like process in A375 melanoma cells*
393
From the nine identified actively transcribed proviruses, five were reported to
express Rec, from which three were also detected in various healthy tissues [28]. To
gain insight into a potential role of Rec in melanoma, we first used an RNAi approach
to deplete Rec expression in two melanoma cell lines (A375 and SKMel-28). A375 and
SKMEL-28 both carry the BRAF^{V600E} mutation, but A375 has a higher proliferation rate,
whereas SKMEL-28 is more invasive [78], indicating their distinct invasion phenotypes.
To knockdown (KD) Rec, we designed three RNAi constructs (**Figure S2**) to target the
sequence of the ERVK6 locus, for which we observed marks of high transcriptional
activity in MITF-High and A375 melanoma cells (**Figure 3**A). Nevertheless, due to the
high level of Rec sequence similarity among several HERV-K proviruses [15], we
predicted our KD constructs would target Rec transcripts produced by most
transcriptionally-competent HERV-K proviruses [15, 28]. In addition, we generated
control cell lines expressing scrambled sequences. In total, we generated 10 stable
cell lines (6 KD and 4 controls) using *Sleeping Beauty* as a stable vector [62, 63]. Real-
time qRT-PCR experiments showed that Rec transcript levels were depleted in each
of the six independent KD cell lines isolated for A375 and SKMel-28 (**Figure 3**D). We
also observed that knocking down Rec also led to an increased level of alternative
HERV-K transcripts encoding Gag and Env in SKMel-28 (all three KD lines) and in
A375 (in the A375-KD1 line only) (**Figure 3**E).
413

To detect transcriptome changes upon Rec KD, we used RNA-seq to compare the

stable KD-Rec lines to their respective control lines. While SKMel-28 expresses Rec

from three genomic loci [43], we were able to unambiguously identify five Rec-

expressing loci in A375 (**Figure 3**F). Next, we computed the differentially expressed

genes (DEGs, |2|-fold and FDR < 0.05) between the Rec KD and their control lines

(**Methods**). We obtained 171 and 773 DEGs upon Rec KD in SKMel-28 and A375

respectively (**Figure 4**A-B**, S1**E**, Table S1,** see **Methods**). In SKMel-28, the 171 DEGs

were not significantly enriched for any Gene Ontology (GO) categories. In contrast, the

773 DEGs in A375 showed significant enrichment for several GO terms including

pigmentation, *cell migration, IL1 production* and *cell proliferation* (FDR < 0.05) (**Figure**

4A).

Notably, we identified a set of genes involved in epithelial-mesenchymal transition

(EMT)-like process, which play a critical role in phenotype switching from the

proliferative to *invasive* state [79-81]. Intriguingly, this set of DEGs included matrix

metallopeptidase 2 (MMP2), a classical marker of cell invasion along with additional

genes previously reported to drive the canonical EMT process [82], such as a zinc

finger protein (SNAI2), forkhead box C2 (FOXC2), cadherin 2 (CDH2) and goosecoid

Homeobox (GSC) (**Figure 4**B). Further, the genes upregulated upon Rec KD in A375

cells included factors such as WNT5B and GSK3B, which are known to induce the

EMT-like process and are activated when melanoma gets relapsed to metastasis after

failed treatment (**Figure 4**B) [83, 84]. During melanoma metastasis, a TF regulatory

network characterized by $SNAI2^{low}/ZEB2^{low}/ZEB1^{high}$ is established [85-88]. The same

TF expression signature is observed upon KD of Rec in A375 cells (**Figure 4**B). Beside

canonical EMT markers, we also observe a strong upregulation of several members of

the SPANX (sperm protein associated with the nucleus in the X chromosome) family

of genes upon Rec KD in A375 cells (**Figure 4**B). The expression of SPANX genes is

normally confined primarily to the germline, but upregulation of SPANXB1/2 has been

previously reported in invasive melanoma [89]. Furthermore, we detected a robust

decrease in the transcript level of MITF(M), the master regulator of the melanocyte

differentiation program, upon Rec KD in A375 cells (**Figure 4**B). In addition to MITF,

we also observed a decreased expression of other melanocyte differentiation markers

(e.g., TYR, PMEL) [90] (**Figure 4**C). Together, the transcriptome changes we observed

upon Rec KD in A375 cells suggests that Rec expression is part of a regulatory
cascade inhibiting the EMT-like process.

To corroborate these results, we chose two of the most DEGs in our transcriptomic
analysis (SPANXB1/2 and MITF(M)) and performed RT-qPCR to compare the mRNA
levels of these two genes in the three independent Rec KD A375 cell lines and their
control lines. The results were consistent with the RNA-seq analysis: SPANXB1/2 was
strongly upregulated (**Figure 4**D) and MITF(M) was robustly downregulated upon Rec
KD (**Figure 4**E).

To further explore the idea that Rec KD in A375 cells recapitulates part of the
expression profile of *invasive* melanoma, we examined the correlation between DEG
upon Rec KD and those differentially expressed between *invasive* and *proliferative*
primary melanoma cultures. We found a significant positive correlation pattern
between the two sets of DEGs (Spearman's rho = 0.20, p-value < 0.05, asymptotic t
approximation) (**Figure 4**F). Overall, these analyses suggest that depletion of Rec in
A375 cells may result in an increased invasion phenotype.

3.4. Depletion of Rec results an enhanced cell invasion in A375 melanoma

468   To test whether Rec expression modulates the invasiveness of melanoma cells,
we performed a trans-well invasion assay to compare the invasiveness of Rec KD
A375 cells to their control lines. The results showed that all three Rec KD lines
exhibited significantly elevated invasiveness (**Figure 5**A-B). These data support the
hypothesis suggested by the transcriptome analyses that Rec may function as a
suppressor of the EMT-like transition process, which modulates the invasiveness of
melanoma.

3.5. Comparison of the Rec KD A375 with the malignant cells in melanoma patients

478   Next, we sought to examine whether a set of dysregulated genes in Rec KD
A375 cells were also differentially expressed in malignant cell types. To assess this,
we first performed a single-cell (sc)RNA-seq analysis (n=4645 cells) derived from

melanoma patients (n=19), which includes both malignant and non-malignant cells

[91]. Consistent with the original study, our analysis distinguishes the *malignant* from

the *non-malignant* cells based on the expression of the most variables genes (**Figure**

6A) and identifies four major *malignant* cell types, marked by MAGEC2, MITF, APOE

(Apolipoprotein E) and SPP1 (Osteopontin) expression, respectively (**Figure 6**A-C). It

is known that high levels of MITF, APOE and SPP1 mark different aspects of

melanoma progression, i.e. melanocyte differentiation, angiogenesis [92] and

inflammation [93], respectively. Interestingly, those three genes were all significantly

downregulated in the Rec KD A375 cells (**Figure 4**B, **7**A), suggesting that the KD of

Rec in those cells recapitulates multiple phases of melanoma.

Having characterized the expression profile of *malignant* and *non-malignant* cells from

melanoma patients, we next compared the ~700 DEGs (**Figure 7**B, see **Methods**)

between *non-malignant* and *malignant* cell types to those detected in our Rec KD A375

experiments (n=773). There were 94 overlapping genes, showing a significant

enrichment over a random expectation (p-value <0.05, chi-square test). Genes

showing malignant-specific expression tend to be upregulated in KD-Rec (p-value =

0.09, Wilcoxon-test, **Figure 7**C).

Lastly, we sought to assess whether genes predicted as downstream targets of MITF

from the same scRNA-seq analysis [91], were also affected in our Rec KD

experiments. We found that 41 out of the top 100 MITF predicted downstream targets

were included in our list of 773 DEGs, a significant enrichment over the random

expectation [p-value, test]. Strikingly, thirty-nine out of the 41 overlapping genes were

downregulated by Rec KD (p-value < $2.2e^{-16}$, chi-square test of observed vs expected).

These observations are consistent with the finding that MITF is downregulated in our

Rec KD lines, thereby affecting MITF downstream targets. (n=41 out of 100, Wilcoxon-

test, p-value < $1.2e^{-11}$) (**Figure 7**D). This analysis suggests that the downstream

targets of MITF might be depleted as the consequence of MITF downregulation upon

KD Rec. Thus, our KD Rec A375 model recapitulates MITF-associated aspects of

phenotype switching in melanoma patients at some degree.

4. Discussion

HERV-K resembles "complex" retroviruses such as HIV-1 by its ability to encode multiple proteins, including Rec. While Env has the typical structure of a transmembrane protein (e.g., extracellular, transmembrane, cytoplasmic domains), Rec lacks the domains required for a membrane protein function (**Figure 8**A). Still, it carries nuclear localization/export signals [8, 19]. Although Env and Rec are different in sequence and function, our results suggest that Rec, like Env [47], exerts a modulatory effect on the EMT-like process in cancer progression. However, the two proteins appear to act in opposite direction. While Env overexpression (both RNA and protein) may promote tumour progression through ERK1/2 activation [47], our KD experiments indicate that Rec may inhibit the EMT process and oppose melanoma metastasis. Interestingly, similar to Env, expression of the Gag protein encoded by HERV-K, also increases the tumorigenic potential of melanoma cells [46]. Because we observed that KD Rec leads to the upregulation of Gag and Env in one of the three KD lines we established using 3 different RNAi constructs, we cannot exclude that some of the transcriptomic and phenotypic (invasiveness) changes we observed are due to indirect effects on Gag and Env levels. Importantly, however, we found that all three KD lines behaved similarly in terms of some of the key differences in gene expression (e.g. MITF(M), compare KD lines in **Figure 4**E) or invasiveness (**Figure 5**). Thus, it appears that Rec's oncogenic potential may oppose that of Env or Gag proteins encoded by HERV-K, and when HERV-K is overexpressed, the balance between these different products may be key to maintain cellular homeostasis.

In about 50% of melanomas, the protein kinase BRAF is mutated (BRAF^{V600E}) [94]. MITF is considered as a crucial factor in melanoma invasiveness [95], and plays a critical role in controlling BRAF^{V600E} melanoma [96, 97]. Accordingly, the expression of MITF has been widely used as a marker to distinguish between the *proliferative* and *invasive* types of melanoma cells [95, 96, 98, 99]. Reduced MITF expression generates *invasive* melanomas, with metastasis-promoting properties [96]. Our data suggest that MITF activates LTR5_Hs in melanoma (not in melanocytes); thus, Rec expression specifically marks the *proliferative* type of melanoma cells. Notably, the *proliferative* phase of melanoma is still a controllable stage of melanoma progression, by contrast to the *invasive* stage. If Rec expression is involved in maintaining a delicate balance between cell proliferation and invasion, as this study suggests, then Rec activation

might be used as a sensitive marker to distinguish between the *proliferative* vs.
invasive phenotypes of melanoma.

Our study indicates that depletion of Rec in BRAF^{V600E} mutant melanoma cells (A375)
results in reduced MITF level. Thus, our data suggest a model whereby a MITF-
dependent regulatory loop is disrupted upon Rec KD through (i) decreased MITF
binding to LTR5_Hs, which leads to (ii) reduced Rec levels and (iii) upregulation of
EMT markers, contributing to (iv) escalated tumour aggressiveness (**Figure 8**B).
Future research is required to decipher the mechanism by which MITF is
downregulated upon Rec KD. It would also be interesting to find out whether Rec
overexpression can lead to enhanced MITF expression in *invasive* melanoma, and
revert the *invasive* phenotype to a *proliferative* one.

Under healthy physiological conditions, apparently a limited number (n=18, or often
less) of HERV-K loci is active [28], whereas, in melanoma patients and/or cell lines,
the number of transcribed HERV-K loci is higher (n~24) [43]. These studies indicate
that there are specific HERV-K loci transcribed under pathological conditions. Although
forced expression of MITF(M) in non-malignant cells has been shown to enhance the
levels of various HERV-K-derived transcripts (e.g., Gag*,* Env, and Rec) [67], the
activated HERV-K loci might differ in both the level of expression and composition of
their HERV-K-derived transcripts. Furthermore, a variable set of HERV-K loci seems
to be expressed in different melanoma patients [43, 100], suggesting that variability in
the amount of HERV-K products may contribute to melanoma heterogeneity [37, 42],
and possibly the progression and outcome of the disease.

Our data suggest that, in addition to the Rec-expressing loci, MITF targets also solo
LTR5_Hss that could function as cis-regulatory sequences (enhancers, promoters) for
nearby genes [64]. Notably, some of the these LTRs harbour the E-box CA(C/T)GTG
motif, which is also recognized by cMYC and MAX oncoproteins [72], suggesting
another connection between HERV-K-derived sequences and oncogenesis. Because
several LTR5_Hs elements are insertionally polymorphic in the human population [2-
5], this type of variation could also have implications in melanoma progression and
deserves future work.

It has been widely discussed how precisely melanoma can be recapitulated in animal models (reviewed in [101]). Interestingly, analysis of single-cell transcriptome data from multiple human patients suggests a distinct separation of *proliferative* and *invasive* types of cells within a patient [5]. However, these two phases do not co-occur in mice [91, 102] As Rec (and HERV-K(HML2)) is hominid-specific, our study might explain certain species-specific features of melanoma progression in humans, and stress the potential limitations of using animal models in melanoma studies.

Figure legends

Figure 1:

A. Heatmap is summarizing the transcription factor (TF) occupancy (ChIP-seq peaks, ENCODE) over individual LTR5_Hs loci in five human cell lines. This plot includes 234 out of 615 LTR5_Hs loci annotated in the human Refseq genome, which is occupied by at least one ChIP-seq peak. Each row represents an individual locus of LTR5_Hs. Each column shows TF ChIP-seq peak occupancy in a different cell line. Various TFs encompassing LTR5_Hs loci are manually annotated in a given cell line. The cluster of TFs binding over a subset of LTR5_Hs loci is shown in the red box.

B. Line plots showing the distribution of normalized ChIP-seq signal counts of cMYC (left panel) and NANOG (right panel) over the annotated LTR5_Hs sequences in the human genome. The comparison is made between primed hESCs (H1) and K562 cells (left panel), as well as primed hESCs (H1) and hESCs (H1), cultured in 3iL "naive" medium (right panel). The ChIP-seq signal counts are calculated relative to primed ESCs, in a 3 Kb genomic window at the left boundaries of LTR5_Hs in the human genome.

C. Heatmap displays the expression of differentially expressed TE families (at counts per million level) upon MITF knockdown in melanocytes (left panel) and melanoma (right panel). (N) denotes the number of differentially expressed TE families (DESeq2 adjusted p-value < 0.05). Note that HERVK expression is affected in melanoma (marked), but not in melanocytes.

D. Venn diagrams illustrating shared and unique differentially expressed TE families between MITF-depleted (KD-MITF) melanocytes and melanoma cells.

615   The upper panel shows down-regulated TEs, whereas the lower panel shows

616   up-regulated TEs upon KD MITF.

617  E. Characterizing primary melanoma cultures according to their MITF expression

618   levels. Bar plots showing the level of MITF expression (Log2 TPM) in primary

619   melanoma cultures (n=10). According to their MITF expression level, cultures

620   are defined as MITF-High (*proliferative*) and MITF-Low (*invasive*) melanoma

621   primary cultures. The annotations indicating the *proliferative* and *invasive* states

622   are taken from the original study [35]. Note that in the following analyses we

623   used the data of MITF-High and MITF-Low as *proliferative* (n=3) and *invasive*

624   (n=2).

626 Figure 2:

627  A. Bar plot showing the LTR5_Hs_HERV-K expression (Log2 CPM) levels in

628   *proliferative* melanoma (MITF-High) and *invasive* melanoma (MITF-Low) cell

629   lines (GSE60664).

630  B. Scatter plot visualizes the average expression (Log2 CPM) (X-axis) and

631   standard deviation from mean (Y-axis) of transposable element (TE) families in

632   various primary melanoma cultures (n=10).

633  C. Line plot showing the MITF ChIP-seq signal occupancy in *proliferative*

634   melanoma (MITF-High) and *invasive* melanoma (MITF-Low) cell lines averaged

635   over all of the LTR5_Hs loci (n=615). Note the E-Box motif, identified in the

636   sequences of LTR5_Hs loci with MITF peaks. The E-Box motif was present in

637   57 out of 115 MITF bound LTR5_Hs sequence in A375 (P-value indicates the

638   probability of the presence of the E-Box motif in MITF bound vs unbound

639   LTR5_Hs loci determined by Chi-square test).

640  D. The effect of PLX4032 treatment on MITF binding at LTR5_Hs loci. Upper

641   panel: Line plot showing the MITF ChIP-seq signal averaged over LTR5_Hs loci

642   in primary melanocytes and melanoma cells with and without the treatment of

643   PLX4032. Lower panel: Heatmap of MITF ChIP-seq signals. Note: MITF

644   occupies LTR5_Hs genomic loci upon PLX4032 treatment, specifically in

645   melanoma (123 unique loci were bound, but not in melanocytes (over 50 loci).

646  E. Bar plot showing the number of MITF peaks detected over the genomic loci of

647   various TE families in melanoma cells treated with PLX4032. This plot shows

648   only those families which have at least five peaks over the individual locus.

Figure 3:

A. Genome browser tracks showing the H3K27Ac, ATAC-seq, and MITF signals at
specific LTR5_Hs-HERV-K sequences at the ERVK6 locus on chr7, encoding
for Gag, Env, and known to produce Rec (UniProt). Note (i) that ERVK6 is a
duplicated, tandemly arranged provirus, flanked by two LTR5_Hs sequences,
mirrored around a centrally located third LTR5_Hs (left and right panels, see also
Figure S2); (ii) that the signals are detectable at the flanking LTRs only; (iii) that
the signals are more robust in both the MITF-High primary melanoma cultures
and A375 cells; (iv) the slight differences in signal intensities over the two
flanking LTR5_Hs. The phylogenetic conservation of the locus is demonstrated
on the middle panel. The schematic structure of the tandemly repeated locus is
shown at the bottom.

B. Normalized ATAC-seq (nucleosome-associated) signals averaged over the
annotated LTR5_Hs sequences in the human genome in A375 cells.

C. Normalized ChIP-seq signals of H3K27Ac (left panel) and H3K4Me1 (right panel)
histone marks (corresponding to active chromatin states) averaged over the
annotated LTR5_Hs sequences in the human genome in A375 cells. Note that
the analysis shows only loci that are mappable from the analysed ChIP-seq data
(likely underestimated).

D. Knocking down (KD) Rec in A375 and SKMel-28 cells using RNA interference.
Stable KD lines were selected in two-rounds by FACS sorting for the GFP-
reporter signal. Bar plots are showing the normalized RT-qPCR quantification of
Rec mRNA levels with three KD-Rec constructs. Control, KD-Scr (scrambled).

E. Knocking down Rec affects the transcription of alternative HERV-K products,
primarily in SKMel-28. Bar plot is showing the effect of knocking down Rec (KD-
Rec) on the expression of alternative HERV-K products, determined by real-time
qPCR. Note: KD-Rec led to an increased level of Gag in SKMel-28 (all the three
biological replicates), whereas Env in A375 (KD1 only). Control, KD-Scr
(scrambled).

F. Five HERV-K loci express Rec in A375. The bar plots demonstrate the locus
specificity of the RNAi strategy against Rec. Relative quantification of Rec
expression (CPM) upon KD at specific genomic loci, compared with scrambled

control (KD1-3 vs. KD-Scr), showing the significantly affected Rec expressing
loci in A375 cells.
684
Figure 4:
A. Significant Gene Ontology (GO) Biological Process terms enrichment analysis
of (differentially expressed genes) DEGs in Rec KD A375 vs. KD-Scr_A375
(scrambled control).
B. Volcano plot showing the comparison of Log2 fold-changes (x-axis) and
adjusted p-value (Log10, y-axis) of DEGs upon KD-Rec vs. KD-Scr in A375
cells. The annotated genes are among the top dysregulated genes that are
known to mark either the *invasive* (red) or the *proliferative* (blue) melanoma.
C. Scatter plot showing the pairwise comparison of DEGs between primary
melanoma cultures and A375 cells. Primary melanoma culture data derive from
the comparison between MITF-high (*proliferative*) and MITF-low (*invasive*) type
of melanomas. A375 data derive from the comparison between KD-Rec and
KD-Scr (scrambled control). The annotated genes are affected in both
comparisons in a similar fashion. Note (i) that MITF-Low (*invasive*) melanoma
and KD-Rec share the upregulation of inflammation-associated genes; (ii) The
downregulated genes are classical markers of *proliferative* melanomas.
D. Validating depleted SPANXB1/B2 expression upon Rec KD in A375 cells vs.
control, using real-time qPCR (previously detected as DEG in RNA-seq
experiments).
E. Validating depleted MITF-M expression (previously detected as DEG in RNA-
seq experiments) upon Rec KD in A375 vs. control, using real-time qPCR (see
Methods).
F. Bar plot representing means ± SEM from RNA-seq expression values (Log2
TPM) for melanocyte markers (e.g., TYR and PMEL) upon Rec-KD in A375 vs.
KD-Scr.
710
Figure 5:
A. Representative inverted light microscope images of trans-well invasion assay
performed in A375 melanoma cells upon knocking-down Rec by three KD
constructs (KD1-3). The Ctrl expresses pT2-CAG-GFP instead of the KD
constructs. Scale bars, 400 µm.

B. Quantification of cells invading in the trans-well invasion assay. ImageJ counted
the number of invaded cells. Columns indicate the average number of invaded
cells per field from two independent experiments. Error bars represent the
means±s.d. of two independent experiments. Mann Whitney test; **$P \leq 0.01$,
****$p \leq 0.0001$ (See **Methods** for details).

Figure 6:

A. UMAP visualisation of the thirteen distinct cell clusters identified by
10XGenomics scRNA-seq in patient-derived melanoma [91]. The identified
clusters are annotated as *Malignant* and *Non-Malignant* (left panel) or according
to their origin (right panel), following original annotations [91]). Note that (i)
despite their proximity to malignant cell clusters, CD248-labelled cells in
Custer9 are most probably *Non-Malignant* cells [103]; (ii) This approach
distinguishes four *Malignant* clusters (e.g., Clusters 10-13).

B. Feature plots based on the UMAP plot shown on (A) visualizing the expression
of selected markers used to identify the distinct cell-types of melanoma. The
Colour intensity gradient indicates the expression of the depicted marker gene.
Each dot represents an individual cell. Dots in green denote lower, whereas in
dark-red, a higher level of gene expression in a given single cell.

C. Violin plots visualize the density and distribution of expression of distinct marker
genes in *Malignant* melanoma clusters (10-13).

Figure 7:

A. Heatmap visualization of scaled (Log2 TPM) expression of SPP1, MITF, APOE
and MAGEC2 upon knockdown (KD) of Rec in A375 cells (three replicates).

B. Heatmap visualization of scaled expression [log TPM (transcripts per million)]
values of distinctive set of ~ 1500 genes (Log2 fold change > 1 and > 50% of
cells expressing either in *Non-malignant* or *Malignant* cells), which are
differentially expressed in scRNA seq data. Colour scheme is based on Z-score
distribution from -2.0 (gold) to 2.0 (purple). Top colour bars highlight
representative gene sets specific to the respective clusters.

C. Box plot showing the comparison of differentially expressed gene sets in
Malignant and *Non-malignant* cells (patient-derived scRNA-seq) and KD-Rec
vs. KD-Scr in A375.

D. Box plot showing the comparison differential expression of a gene set of "MITF-
High" (consisting of MITF itself and the top 100 genes correlated with MITF
expression) between *Malignant* and *Non-malignant* cells (patient-derived
scRNA-seq) and KD-Rec vs. KD-Scr in A375.

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Figure 8:
A. Schematic structure of the HERV-K-derived alternative splice products, Env and
Rec, and their effect on the endothelial-mesenchymal transition (EMT)-like
process of cancer progression (EMT(L). The signal peptide (SP);
transmembrane subunit (TM); Splice acceptor (Sa); Splice donor (Sd).
B. Modeling the interaction loop between MITF and LTR5_Hs and Rec in
melanoma. MITF binds to LTR5_Hs and drives Rec expression from certain
Type 2 proviruses. Rec marks the proliferative stage melanoma cells and,
similarly to MITF, inhibits the endothelial-mesenchymal transition (EMT)-like
process (EMT(L). Depletion of Rec results in MITF downregulation and the
failure to maintain the balance between cell proliferation and invasion.

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Figure S1:
A. (Left panel) Ranked correlation matrix of genes and 42 transposable elements
(TE) families, which are differentially expressed between *proliferative* (MITF-
High) and *invasive* (MITF-Low) primary melanoma cultures. Genes and TE
families are manually annotated. Colour codes on the left side of the heat map
panel represent the unsupervised clusters of most variable genes and TEs
across the samples. On the right, selected genes are listed that are positively
or negatively correlated with LTR5_Hs expression across the melanoma
primary cultures.
B. Distribution of ChIP-seq signal of the H3K4Me3 histone mark occupancy,
corresponding to active promoters averaged over LTR5_Hs loci in MITF-High
(*proliferative*) vs. MITF-Low (*invasive*) state of primary melanoma cultures.
C. Distribution of ChIP-seq signal of the H3K27Ac histone mark occupancy (lower
panel) averaged over LTR5_Hs loci in MITF-High (*proliferative*) vs. MITF-Low
(*invasive*) state of primary melanoma cultures.
D. Heatmap displays the DEGs upon KD-Rec in A375 cells vs. KD-Scr.

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784 Figure S2:

Schematic of the positions targeted by the knockdown (KD1,2,3) constructs on the ERVK6(HML2.HOM) locus. ERVK6 is a tandemly organized HERV-K(HML2) provirus, possessing three LTRs. The genomic coordinates of the locus on chr seven are indicated. Both are type 2 and encode identical Rec proteins (UniProt). A human-specific provirus encodes these Rec proteins. Note that due to the high conservation of Rec sequences [15], the KD was targeted multiple loci in the human genome.

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1082

A Encode ChIP-seq (N=142)

LTR5_Hs loci (N=234)

K562
H1ESC
HepG2
GM12878
HUVEC
Peak/No Peak

cMYC, MAX, JUND, PU.1, P300

B

cMYC

Log2 CPM+1

K562
hESCs

Distance from LTR5_Hs left boundary (Kb)

NANOG

Log2 CPM+1

3iL Naive
hESCs

Distance from LTR5_Hs left boundary (Kb)

C

N=12

MER41D
LTR23-int
MLT1G3
LTR39
HERVE_a-int
HERVL66-int
LTR26
MER52-int
MER72B
LTR59
MER85
MER57B1

N=08

MLT1F1-int
LTR3A
MER57A-int
LTR40A1
MER63A
HERVK14-int
SVA_F
MER67C

Control Control Control KD-MITF KD-MITF KD-MITF

HeMa Melanocyte

N=17

LTR3A
HERVI-int
HERVE_a-int
LTR10D
HERVK-int ◄
HERVK3-int
MER52-int
HERVH-int
MLT1G

N=68

Scaled Log2 FPKM

KD-MITF KD-MITF Control Control

501 Melanoma

D

Downregulated TE families in KD-MITF

14 03 09

MER42-int
MER72-int
HERVE_a-int

Upregulated TE families in KD-MITF

66 02 06

SVA-F
LTR40A1

Melanoma
Melanocyte

E

Primary Melanoma cultures (N=10)

Proliferative

Invasive

MITF expression (Log2 TPM)

MITF-High
MITF-Low

A

Ctrl	KD1-Rec

KD2-Rec	KD3-Rec

B

A — Scaled Log2 TPM

B — Cell-types / Expression

- CD8 T-cells
- Naive CD4 T-cells
- CD4 T-cells
- NK cells
- B-cells
- Endothelial
- Ki67 Senescent
- CD248 cells
- MITF/MAGEC2 Melanoma
- MAGEC2/SPP1 Melanoma
- MITF Melanoma
- APOE Melanoma

A

B

A

pairwise ranked correlation
~ 3200 genes and 42 TE families

-1 -0.5 0 0.5 1

Invasive network

TGFBI
GSC
SNAI1
ID1
ID2
MMP2
WNT5A
FOXC2

Proliferative network

MITF
SNAI2
ZEB2
SPP1
LTR5_Hs
HERV-K
LTR2C
LTR13

B

positively correlated genes (n=72)

MITF
rho = 0.43, p-value < 0.2e-9

negatively correlated genes (n=66)

-0.5 0 0.5 1

Level of Ranked correlation with LTR5_Hs

cor.test, p-value < 1e^{-5}

C

H3K4Me3

Log2 counts per million

0.75
0.50
0.25
0

LTR5_Hs

-2 -1 0 1 2

Distance from left boundary

Proliferative

Invasive

D

H3K27Ac

Log2 counts per million

3
2
1
0

LTR5_Hs

-2 -1 0 1 2

Distance from left boundary

E

Scaled expression
Log2 TPM

-2 -1 0 1 2

Control1 Control2 Control3 KD1-Rec KD2-Rec KD3-Rec