# Co-expression of Bovine leukemia virus and Bovine foamy virus-derived miRNAs

in naturally infected cattle

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## Abstract

Among the viruses that encode miRNAs and infect cattle such as Bovine leukemia virus (BLV), Bovine foamy virus (BFV) and Bovine herpesviruses (BoHV), BLV has gained attention due to the critical role that BLV-miRNAs may play in inducing lymphosarcoma in infected animals. BLV is highly prevalent in the Americas and negatively affects dairy herds, primarily due to restrictions on the commercialization of dairy products from infected animals and a decrease in milk production. Mixed infections involving BLV and BFV appear to be common in cattle. Considering the ability of foamy viruses to cross species barriers, preventing their presence within the food chain is essential. We identified the co-expression of seven BLV-derived miRNAs (blv-miR-B1-3p, blv-miR-B2-5p, blv-miR-B2-3p, blv-miR-B3-5p, blv-miR-B3-3p, blv-miR-B4-3p and blv-miR-B5-5p) and three BFV-derived (bfv-miR-BF1-5p, bfv-miR-BF1-3p, and bfv-miR-BF2-5p) in naturally BLV-infected cows. Besides, seven differentially expressed  bovine miRNAs (bta-miR-375, bta-miR-133a, bta-miR-677, bta-miR-1, bta-miR-3613a, bta-miR-9-5p and  bta-miR-95) were identified between cows with high BLV proviral load and uninfected counterparts (fold change > $|1.5|$ and q-value < 0.05). A comprehensive functional analysis of protein-protein interaction networks for genes targeted by both viral and host-derived miRNAs highlighted key pathways implicated in tumorigenesis and immune response. Although BLV and BFV derived miRNAs target different genes, their functional convergence may reflect coordinated viral modulation of host cellular processes, raising important concerns regarding their potential influence on disease severity and the increased dissemination of BFV. These findings offer new perspectives for creating diagnostic and treatment approaches to manage viral persistence and tumorigenesis in cattle.

## Importance

The Bovine leukemia virus (BLV) and Bovine foamy virus (BFV) are retroviruses that encode miRNAs and infect cattle. While the role of BFV-derived miRNAs remains unclear, BLV-miRNAs have gained attention for their potential involvement in oncogenesis. Mixed BLV-BFV infections are common and given foamy viruses' potential to cross species barriers, it is essential to prevent their presence in the food chain. We reported the co-expression of three BFV-derived miRNAs and seven BLV-derived

miRNAs in naturally infected cattle. A functional analysis of the protein-protein interaction graph of genes

potentially targeted by both viral and host-derived miRNAs revealed key metabolic pathways associated

with tumorigenesis and immune response regulation. The co-expression of BLV and BFV miRNAs

suggests a potential functional interactions that may influence disease progression and BFV

dissemination. These findings offer opportunities for developing diagnostic and therapeutic strategies to

control viral persistence and tumor development in cattle.

## Introduction

The study of virus-host interactions has traditionally focused on how viral proteins modulate host proteins

to hijack cellular processes in favor of viral replication and pathogenesis (1) However, like their host

cells, many viruses encode regulatory non-coding RNAs, including small RNAs such as microRNAs

(miRNAs), which provide new capacities for regulating host gene expression to ensure successful viral

infection (2; 3; 4; 5). miRNAs are enzymatically processed small RNAs, typically 19–24 nucleotides (nt)

long, that fine-tune gene expression at the post-transcriptional level (6). The regulatory function of

miRNAs is primarily mediated by the RNA-induced silencing complex (RISC), which facilitates base-

pairing between the mature miRNA´s 2–7 nucleotide seed region and the 3' untranslated region (3'UTR)

of the target messenger RNA (mRNA) (7; 8). The miRNA-mRNA interaction induces mRNA degradation

by cellular endonucleases or translational repression (8; 9). It is estimated that miRNAs may regulate at

least 60% of the human coding genome (10; 11).

Host-derived microRNAs (miRNAs) regulate viral infections by exhibiting a dual nature that can benefit

either the virus or the host. For example, host miRNAs that directly target viral RNA often enhance the

viral life cycle. Conversely, host miRNAs can exert an indirect negative effect on the virus by targeting

mRNAs that encode essential host factors necessary for the viral cycle or for establishing immune

responses and defense mechanisms (5; 12). Given their high versatility in fine-tuning gene expression

and their small coding size, it is not surprising that viruses have evolved the ability to express their own

miRNAs (13). Since the first report of miRNAs encoded by the Epstein-Barr virus (EBV) (14), miRNAs

encoded by DNA viruses have proven to be the most abundant, particularly those expressed by the

Herpesviridae family. In contrast, RNA viruses that express miRNAs are far less common and are

typically restricted to those with a DNA intermediate. Additionally, several miRNA-like molecules have

been identified in both positive and negative-sense RNA viruses (3; 15; 16; 17; 18; 19; 20; 21; 22; 23;

24; 25; 26; 27; 28).

Usually, viral miRNAs follow the same biogenesis and effector pathways as host miRNAs. However,

certain retroviruses, such as members of the Spumaretrovirinae subfamily (e.g., Simian foamy virus

(SFV) and Bovine foamy virus (BFV)) and the Orthoretrovirinae subfamily (e.g., Bovine leukemia virus

(BLV)), decouples miRNA transcription from genomic transcription by employing RNA polymerase III

instead of RNA polymerase II (21, 22; 23). This mechanism likely prevents the auto-degradation of viral

genomes mediated by endonucleases during miRNA biogenesis. (29; 30).

Generally, the functional role of viral miRNAs involves the downregulation of both viral and host proteins;

however, upregulation would also occur—either indirectly through the modulation of gene that act as

transcriptional repressors, or directly, as some miRNAs can function as transcriptional activators by

binding to specific DNA sequences or interacting with factors that enhance gene expression (18; 31; 32).

Collectively, these actions modulate gene expression to regulate the viral life cycle (e.g., promoting or

inhibiting cell proliferation and apoptosis), evade the host immune response, and establish latent or

persistent infections (4; 16; 33; 34; 35). This is particularly evident in retroviruses and herpesviruses,

where viral protein expression is strongly restricted during latency. The production of a low-antigenicity

regulatory molecule provides a significant advantage for viral persistence while evading host immune

detection (13; 36).

Viruses that encode miRNAs and infect cattle include BLV (21), BFV (23), Bovine herpesvirus 1 (BoHV-

1) (37), and Bovine herpesvirus 5 (BoHV-5) (38). BoHV-1 encodes at least ten miRNAs, two of which

have been implicated in maintaining latency both *in vivo and in vitro* (39; 40). In contrast, the functional

roles and target genes of the three BFV-derived miRNAs remain poorly understood (41). Among BLV-

encoded miRNAs, blv-miR-B4-3p is the most extensevely studied. It shares a seed sequence with miR-

29a, an oncomiR that targets tumor suppressors such as peroxidasin homolog (*PXDN*) and HMG-box

transcription factor 1 (*HBP1*), as reported in murine tumor-induction studies (42; 43). *In vitro* reporter

assays have demonstrated that blv-miR-B4-3p exerts a repressive effect on the expression of *HBP1* and

*PXDN*. However, a negative correlation between *blv-miR-B4-3p* and *HBP1* expression has been

documented, while levels of *PXDN* remained unchanged in ovine primary tumor B cells (21; 44).

Additionally, recent findings indicate that *PXDN* expression is significantly downregulated in naturally

BLV-infected cattle expressing blv-miR-B4-3p when compared to their non-infected counterparts.

Notably, miR-29a expression remains unaffected in both groups (45).

The BLV prevalence reaches up to 85% in North America at farms-level and over 80% in lactating cows

of Argentina's main dairy region (46; 47; 48; 49). International trade restrictions of livestock products

from affected herds impact negatively the economy (50), but greater losses stem from reduced milk

production and earlier culling of asymptomatic BLV carriers compared to BLV-free herds (51; 52).

Mixed infections involving BLV and BFV appear to be common in cattle (53). Given the ability of foamy

viruses (FVs) to cross species barriers and cause zoonotic infections in humans, their presence in the

food chain may be a potential risk to both human and animal health. Furthermore, cattle infected with

BLV are more susceptible to co-infections with BoHV-1 (54; 55; 56).

This study identified the expression of three BFV-derived miRNAs and seven BLV-derived miRNAs in

naturally BLV-infected cattle through small RNA sequencing analysis. No miRNAs from BoHV-1/5 were

found. Seven bovine miRNAs showed differential expression between cattle infected with BLV/BFV and

those uninfected with BLV. A functional analysis of the protein-protein interaction network for genes

potentially targeted by both viral and host-derived miRNAs revealed key metabolic pathways associated

with tumorigenesis and immune response regulation.

**Methods**

**Selection of Animals**

BLV-infected and non-infected samples were obtained from a previously phenotyped population as

described in Petersen et al. (2021) (57). Briefly, 129 adult Holstein cows (over 3 years old sharing the

same lactation period) from a dairy farm in the central region of Argentina -where the average individual

prevalence of BLV exceeds 80% (49)- were initially screened. Since antibody levels have been reported

to reflect proviral load (PVL) *in vivo* (49, 58), anti-BLV ELISA (see below) was assesed at −10 months

(T1) and −5 months (T2) before final sampling; mean percentage of reactivity (PR) was 122.7 ± 34.8 at

T1 and 146.3 ± 55.6 at T2. Animals in the highest PR quartile (Q) (Q4: T1 = 148.6–178.6%, T2 = 194.4–

239.6%) and those testing negative at both times were selected for further proviral load (PVL)

quantification via qPCR (details provided later) at −3 months (T3) and at the time of sampling (T4).

Based on the consistency of PVL results, four cows with persistently high PVL (HPVL) and three cows

with consistently negative qPCR results were ultimately selected for small RNA sequencing. All sample

collection and handling procedures adhere to the recommendations and guidelines of the Institutional

Committee for the Care and Use of Experimental Animals at the Instituto Nacional de Tecnología

Agropecuaria.

**Isolation of Peripheral Blood Mononuclear Cells (PBMCs)**

Fresh blood samples from animals were collected via jugular venipuncture and supplemented with EDTA

(225 μM). Peripheral blood mononuclear cells (PBMCs) were isolated on the same day of collection

using Ficoll-Paque Plus (GE Healthcare, Uppsala, Sweden) density gradient centrifugation, following the

manufacturer's protocol. The plasma fraction was kept for anti-BLV ELISA serology. After isolation,

PBMCs were preserved in RNAlater solution (Ambion, Austin, TX) and stored at −80°C until further use.

**Serology**

The anti-BLV ELISA assay, as described by Trono et al. (2001) (59), was employed to identify BLV-

infected animals. The whole BLV viral particle served as the antigen for detecting plasma anti-BLV

antibodies in each sample. Briefly, a sample-to-positive (S/P) ratio, referred to as the percentage of

reactivity (PR), was calculated using the following formula: $S/P = [(OD_{Sample} - OD_{WS-}) / (OD_{WS+1/7} - OD_{WS-})]$

× 100, where OD refers to optical density, WS- indicates the negative serum, and WS+1/7 is the

international standard weak positive control serum (diluted 1:7 in negative serum). Samples with a PR

greater than 25% were considered positive (BLV(+)), while those below were regarded as negative

(BLV(-)).

**BLV proviral load (PVL) Quantification**

Genomic DNA was extracted from PBMCs using the High Pure PCR Template Preparation Kit (Roche,

Penzberg, Germany) following the manufacturer's instructions. The quality and concentration of genomic

DNA from whole blood samples, extracted with the Blood Genomic DNA AxyPrep™ kit (Axygen

Biosciences, Union City, USA), were assessed using a microvolume spectrophotometer (NanoDrop™

Technologies, Inc., Wilmington, USA).

A BLV POL gene-based PVL qPCR assay based on the SYBR Green dye detection system was

conducted as described by Petersen et al. (2018) (60). Each 25 µL qPCR reaction contained Fast Start

Universal SYBR Green Master Mix (2×; Roche), forward and reverse primers (800 nM; BLVpol_5f:5′-

CCTCAATTCCCTTTAAACTA-3′ and BLVpol_3r:5′-GTACCGGGAAGACTGGATTA-3′; Thermo Fisher

Scientific), and 200 ng of genomic DNA template. Amplification and detection were carried out using a

Step One Plus system (Applied Biosystems, Foster City, CA). The specificity of each BLV-positive

reaction was confirmed by melting temperature dissociation curve (Tm) analysis. Based on the

assumption of a low natural infection rate (1% of BLV-infected cells in the peripheral blood), PVL values

below 1,500 copies/µg of total DNA were classified as low, while values above this threshold were

classified as high (61).

**Small RNA Sequencing**

Total RNA was extracted from PBMCs using the miRNeasy kit (Qiagen, Hilden, Germany) according to

the manufacturer's protocol. RNA quality (integrity) and concentration were assessed via digital

electrophoresis on an Agilent 2200 TapeStation system (Agilent Technologies, Santa Clara, US).

The QIAseq miRNA Library Kit (Qiagen, Hilden, Germany) was used for miRNA sequencing library

construction, which integrates Unique Molecular Indices (UMIs) tags into the adapters during the cDNA

synthesis step. These UMI tags help to mitigate PCR amplification biases and sequencing artifacts by

allowing identification and collapsing reads with the same amplification origin. A specific 8-nt barcode

was assigned to each sample for multiplexing. The multiplexed libraries were pooled to an equal molar

concentration and processed in a NovaSeq run (Illumina, San Diego, US) with a 75x50 bp configuration

(300–400 million reads pairs per lane).

**Small RNA sequencing reads processing and deduplication**

The quality control of sequencing reads was performed sequentially to ensure reliable reads for

unbiased miRNome identification and quantification. The 12-nt UMI pattern was extracted from each

read using the umi_tools v1.1.6 software (62). Forward (Fw) and reverse (Rv) reads were processed

independently before collapsing. Reads with low quality (quality cutoff < 30), a minimum length < 16 nt,

the presence of ambiguous bases (N) or adapter sequences were removed using Cutadapt v5.0 (63).

Singleton reads resulting were discarded, and only paired Fw and Rv reads were considered. Forward

read sequences were aligned to the Rv-complement and merged into a single consensus read using

BBMap v25.85 (64) with a minimum insert size of 16 nt and a minimum overlap of 18 nt. After merging,

Qiagen-specific adapter sequences were removed using Cutadapt v5.0 (63).

A stringent mapping strategy was employed for read deduplication using Bowtie2 v2.5.4 (65). Reads

were aligned to the bovine reference genome ARS-UCD1.2 (acc. GCF_002263795.1) and to viral

genomes encoding miRNAs that infect cattle, including BLV (acc. NC_001414.1), BFV (acc.

NC_001831.1), BoHV-1 (acc. NC_063268.1) and BoHV-5 (acc. NC_005261.3). No mismatches were

allowed in the first 18 bases of the read (left end). After alignment to the reference sequences, reads

were deduplicated based on their mapping coordinates and the previously extracted UMI tags using

UMICollapse v1.0.0 (66).

**miRNAs identification and quantification**

Deduplicated reads were mapped to the reference precursor miRNA sequences for bovine, BLV, BFV,

BoHV-1 and BoHV-5 obtained from miRBase v22.1 (67). Mature miRNA expression was quantified using

miRDeep2 v0.1.3 (68) with default parameters, generating a count matrix of miRNA expression for each

sample.

**Differential miRNA Expression analysis**

The raw count matrix for all samples was analyzed using DESeq2 v1.44.0 (69) to assess differential

miRNA expression between BLV-infected and non-infected animals. Based on miRNA expression

profiles, an exploratory analysis of variation patterns between infected and non-infected animals was

performed using principal component analysis (PCA). To ensure homoskedasticity of the data, raw

miRNA counts were transformed using the Variance Stabilizing Transformation (VST) method

implemented in DESeq2 v1.44.0. Pairwise distances between samples were calculated based on the

root-mean-square deviation of miRNA expression levels.

To test for differential miRNA expression, the DESeq2 model adjusts miRNA counts by accounting for

sample-specific differences, such as library size, sequencing composition, and miRNA-specific biases,

using negative binomial models. Under the null hypothesis, the model assumes that a miRNA has the

same mean expression across the two conditions (BLV(-) and BLV(+) groups). The statistical

significance of the $Log_2$ fold change ($Log_2FC$) between conditions was assessed using the Wald test.

Correction for multiple comparisons was performed using the Benjamini-Hochberg false discovery rate

(BH-FDR) method (70). The miRNAs with BH-FDR-adjusted p-values (q) < 0.05 and a $Log_2FC > |1.5|$

were considered statistically significant. The PCA and Volcano plots were performed under the R

environment (71).

**miRNAs gene target prediction**

Significant differentially expressed bovine and viral-derived miRNAs between BLV(+) and non-infected

cows were used to predict potential bovine gene targets *in silico* (miRNA:mRNA interactions) using

miRanda v3.3 (72), PITA v6.0 (73), and RNAhybrid v2.1.1 (74), with default parameters. The analysis

utilized 3´-UTR sequences of 21,400 bovine genes from the ARS-UCD1.2 annotation (75). Each gene

target prediction algorithm employed a distinct strategy: miRanda relied on seed-site alignment and the

thermodynamic stability of RNA-RNA duplexes; PITA incorporated target-site accessibility; and

RNAhybrid uses the minimum free energy of hybridization between miRNA-mRNA sequences. Predicted

targets were filtered using the following threshold: a score of ≥140 and free energy ≤−20 kcal/mol

(miRanda) and free energy ≤−20 kcal/mol (PITA and RNAhybrid). Only gene targets predicted by all

three tools (consensus gene targets) were retained for downstream analysis

**Gene ontology and graph-based pathway analysis of miRNA target genes**

The functional annotation of the potential target genes was carried out using STRINGdb (76) with Gene

Ontology (GO) terms and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways (77). Next, a

protein-protein interaction (PPI) network of the identified target genes was constructed using STRINGdb.

This platform integrates curated and predicted interactions from multiple sources, including experimental

evidence, computational predictions, scientific literature, co-expression data and primary public

databases. Highly interconnected protein clusters within the network were identified using the Markov

Cluster Algorithm (MCL) with default parameters (inflation parameter = 2.5). Only clusters with more than

15 nodes (genes) were considered. GO terms and KEGG pathways overrepresentation analysis was

performed on the identified clusters using Fisher's exact test based on the hypergeometric distribution,

using a term similarity threshold > 0.7 (78). Wordclouds plots of significant GO terms grouped by

semantic similarity were obtained using the rrvgo package with default parameters (79). A GO term

category–gene target graph was generated to visualize the connections between significantly enriched

GO terms and their associated genes using the R graphical environment (71).

## Results

In this study, four animals consistently exhibited both high antibody reactivity across all sampling times

(IND_5671 = 178.6 ± 28.4; IND_5841 = 191.3 ± 32.8; IND_6021 = 169.8 ± 42.7 and IND_6097 = 183.2 ±

25.4) and high proviral loads at T-3 and T0 (average proviral load: IND_5671 = 63,148.0; IND_5841 =

44,234.3; IND_6021 = 52,902.4 and IND_6097 = 91,255.4). These cows were classified as BLV(+). In

contrast, three animals that consistently tested negative for both anti-BLV ELISA and qPCR were

classified as BLV(-).

Total RNA quality was confirmed (RIN 7.8–9.6), and small RNA libraries were prepared using UMI-

tagged adapters. High-quality, non-redundant reads were obtained through a multi-step processing

pipeline. A descriptive characterization of each sample, the number of sequencing reads before and after

quality control; merging and deduplication; and mapping statistics are summarized in Table 1.

Demultiplexed reads were mapped to the reference miRNA sequences of bovine, BLV, BFV, BoHV-1

and BoHV-5 (n = 1060) to identify and quantify the miRNAs present in each sample. On average, 351

miRNAs (with mapped reads ≥ 5) derived from both bovine and viral sources were identified and

quantified across all samples (Supp. Table S1). A multivariate analysis of sample distances based on the

normalized miRNA expression matrix revealed that BLV(-) samples clustered together, forming a distinct

group separate from BLV(+) samples. PC1 explains 62% of the variance in the expression matrix, while

PC2 accounts for 17% (Figure 1).

In the BLV(+) group, a total of 10 viral-derived miRNAs were consistently expressed, including 7 from

BLV: blv-miR-B1-3p (avgexp = 27,681.1), blv-miR-B2-5p (avgexp = 59,058.0), blv-miR-B2-3p (avgexp =

112.5), blv-miR-B3-5p (avgexp = 309.6), blv-miR-B3-3p (avgexp = 49,312.2), blv-miR-B4-3p (avgexp =

65.1), and blv-miR-B5-5p (avgexp = 951.7). Additionally, 3 BFV-derived miRNAs were identified: *bfv-*

*miR-BF1-5p* (avgexp = 6,844.7), bfv-miR-BF1-3p (avgexp = 8,257.3), and bfv-miR-BF2-5p (avgexp =

14,920.6). In contrast, viral-derived miRNAs expression was low (avgexp < 30.0) or undetectable in the

BLV(-) group, except for blv-miR-B2-5p (avgexp = 67.4) and blv-miR-B3-3p (avgexp = 49.8) (Table 2).

The mapping of all reads from BLV(-) samples to the bovine reference genome (acc. ARS-UCD1.2)

showed that only 716 out of 14,378,553 reads failed to align, indicating that nearly all reads were derived

from bovine small RNAs. Besides, the number of reads aligned to the BLV (acc. NC_001414.1) and BFV

(acc. NC_001831.1) reference genomes from BLV(-) cows were only 600 and 141 reads, respectively

(Table 1). Moreover, the alignment of the same reads dataset to the identified BLV- and BFV- miRNAs

resulted in 642 reads with at least one aligment. Among these, only six reads also map to the bovine

reference genome, and none aligned any other known miRNA from any organism in the miRBase

database (n = 48,871 mature miRNAs). The Supp. Figure S2 shows the coverage of all reads from

BLV(+) and BLV(-) cows across the reference genomes of BLV and BFV,  emphasizing that detectable

read coverage is restricted to known miRNA regions for both viral genomes.

Differential expression analysis of host miRNAs, identified a total of seven bovine miRNAs (bta) with a

fold change (FC) > |1.5| and a q-value < 0.05 as differentially expressed (BTA-miRNAs-DE) between the

BLV(+) and BLV(-) groups: bta-miR-375, bta-miR-133a, bta-miR-677, bta-miR-1, bta-miR-3613a, bta-

miR-9-5p, and bta-miR-95 (Figure 2; Supp. Table S2). In Supp. Figure S3, a heatmap was generated to

visualize the expression of miRNAs in BLV(+) and BLV(-) cows. A hierarchical clustering was applied to

both dimensions to show groups of samples with similar expression profiles, as well as clusters of

miRNAs with coordinated expression patterns.

To evaluate the potential functional impact of the seven differentially expressed BTA-miRNAs and the

ten viral-derived miRNAs, a target gene prediction analysis was performed. A total of 1,518 genes were

identified as putative targets, based on the consensus of three independent prediction tools. Among

these, 281, 977, and 260 genes were exclusively targeted by BTA-miRNAs-DE, BLV-miRNAs, and BFV-

miRNAs, respectively (Supp. Table S3). No overlap was observed among the predicted target. Then,

functional annotation of the 1,518 potential target genes was conducted using Gene Ontology (GO)

terms and metabolic pathways, revealing 91  significantly overrepresented GO biological process (BP)

terms, 10 GO molecular function (MF) terms and 30 GO cellular component (CC) terms. The word cloud

visualizations for the enriched GO terms in the GO BP and GO MF categories are shown in Figure 3, the

most frequent occurring words for the overrepresented processes and functions included "Regulation",

"Process", "Response", "Binding", "Activity", among others. The overrepresentation analysis of KEGG

and Reactome metabolic pathways revealed four significant terms: "*Ras signaling pathway*" (bta04014, q

= 0.04), "*Pathways in cancer*" (bta05200, q = 0.02), "*Oxytocin signaling pathway*" (bta04921, q = 0.024),

and "*Cushing syndrome*" (bta04934, q = 0.04). Additionally, the terms "*Signal Transduction*" (BTA-

162582, q = 0.0002) and "*Metabolism*" (BTA-1430728, q = 0.01) were significant in Reactome pathways

analysis (Supp. Table S4).

A protein-protein interaction (PPI) network was developed, comprising 1,518 nodes (genes) and 6,075

edges (interactions) (Supp. Figure S2). From this network, three highly interconnected protein clusters

(C1, C2, and C3) were identified (PPI p-value < $1.0e^{-16}$), which potentially represent hubs of specialized

biological function (Figure 4). Functional overrepresentation analysis for these clusters revealed key

cellular processes involved in immune response modulation, cell signaling, and mechanisms related to

cancer development (Supp. Table S4). Interestingly, despite the lack of overlap among target genes of

BTA-, BLV-, and BFV-derived miRNAs, their predicted functions converge on interconnected hubs,

indicating potential interplay at the pathway level. The enrichment graph plot depicts the interconnections

among the top 15 enriched BP GO terms per cluster, highlighting functional overlap through target genes

(Figure 5).

For example, significantly overrepresented GO BP terms for C1 (q < 0.05), included "*Cytokine-mediated*

*signaling pathway*", "*Signal transduction*", "*Positive regulation of leukocyte chemotaxis*", "*Positive*

*regulation of immune system process*", "*Regulation of leukocyte proliferation*", "*Regulation of cell*

*adhesion*", "*Myeloid leukocyte migration*", "*Positive regulation of T cell activation*", "*Regulation of*

*leukocyte apoptotic process*", "*Regulation of interferon-gamma production*", amog others, focusing

especially on processes related to immune activation, cytokine production, and immune regulation. For

C2, key terms represent biological processes primarily tied to cell growth, development, and tissue

morphogenesis, such as "*Transmembrane receptor protein tyrosine kinase signaling pathway*", "*Vascular*

*endothelial growth factor signaling pathway*", "*ERBB2-EGFR signaling pathway*", "*Positive regulation of*

*MAP kinase activity*", "*Hemopoiesis*", "*Negative regulation of apoptotic process*", "*Anatomical structure*

*morphogenesis*", "*Positive regulation of epithelial cell proliferation*", "*Positive regulation of mast cell*

*chemotaxis*", among others.

In C3, the terms are mostly associated with epigenetic control and transcriptional regulation through

chromatin modification, including "*Histone deacetylation*", "*Histone modification*", "*Chromatin*

*remodeling*", "*Negative regulation of transcription by RNA polymerase II*", "*Chromosome organization*",

"*Regulation of apoptotic signaling pathway*", among others.

Significantly enriched KEGG and Reactome pathways for clusters included, C1: "*Cytokine-cytokine*

*receptor interaction*" (bta04060, q-value = $8.40e^{-22}$), "*Viral protein interaction with cytokine and cytokine*

*receptor*" (bta04061, q-value = $1.02e^{-16}$), "*Cytokine Signaling in Immune system*" (BTA-1280215, q-value

= $1.09e^{-06}$), "*Chemokine receptors bind chemokines*" (BTA-380108, q-value = $1.10e^{-06}$), "*Interleukin-21*

*signaling*" (BTA-9020958 q-value = 0.0057), "*IL-6-type cytokine receptor ligand interactions*" (BTA-

6788467, q-value = 0.0105); C2: "*MAPK signaling pathway*" (bta0401, q-value = $3.50e^{-10}$), "*Regulation of*

*actin cytoskeleton*"  (bta04810, q-value = 0.0014), "*HIF-1 signaling pathway*" (bta04066, q-value =

0.0029), "*Signaling by Receptor Tyrosine Kinases*" (BTA-9006934, q-value = $3.13e^{-16}$), "*PI5P, PP2A and*

*IER3 Regulate PI3K/AKT Signaling*" (BTA-6811558, q-value = $1.62e^{-09}$), "*PI3K Cascade*" (BTA-109704,

q-value = 0.0034), "*Regulation of KIT signaling*" (BTA-1433559, q-value = 0.0058); and C3: "*Regulation*

*of TP53 Activity through Acetylation*" (BTA-6804758, q-value = $2.85e^{-08}$), "*RNA Polymerase I*

*Transcription Initiation*" (BTA-73762, q-value = $1.02e^{-07}$), "*Gene expression (Transcription)*" (BTA-74160,

q-value = $3.05e^{-05}$), "*Chromatin modifying enzymes*" (BTA-3247509, q-value = $4.15e^{-05}$), "*Regulation of*

*PTEN gene transcription*" (BTA-8943724, q-value = 0.00052).

## Discussion

In this study, we performed a comprehensive characterization of circulating miRNAs in the peripheral blood of naturally BLV-infected and non-infected cattle. Notably, all BLV-infected cows included in the analysis consistently exhibited HPVL levels across time points; a condition previously associated with BLV pathogenesis progression (80; 81) and increased risk of viral transmission within herds (82). Following the identification of BLV-miRNAs in persistently infected cell lines (21), their expression was assessed using next-generation sequencing for complete miRNA profiling and RT-qPCR to test the expression of candidate miRNAs. BLV-miRNAs were identified in primary leukemic B-cells and B-cell lymphomas isolated from BLV-infected ovine/bovine tumors (44), experimentally infected cattle (83), and naturally BLV-infected cattle (45; 84; 85).

On the other hand, BFV-miRNAs have been identified in both persistently and recently infected MDBK cells, as well as in BFV experimentally challenged cattle (23). However, to our knowledge, this is the first report of the co-expression of miRNAs derived from two distinct viruses (BLV and BFV) in naturally infected cattle. We identified and quantified seven BLV-derived and three BFV-derived miRNAs. No miRNAs from BoHV-1 and -5 were detected.

Coinfections of BLV and BFV are prevalent among cattle (53), akin to the occurrences in cats infected with Feline foamy virus (FFV) and Feline leukemia virus (FeLV) (86), as well as in baboons with Simian foamy virus (SFV) and Simian T-cell leukemia virus (STLV) (87). While BFV infections in cattle are generally regarded as mild and asymptomatic (23; 53), there is evidence to suggest that co-infections with BFV and BLV could amplify pathogenicity and impair the bovine immune system, thereby aiding in the transmission and spread of BFV (53; 88; 89; 90). Infectious BFV has been extracted from raw milk, and considering that foamy viruses can cross species barriers, there is rising concern over the zoonotic potential of BFV in humans (91; 92). Moreover, increased pathogenic effects have been noted in mixed infections of macaques with SFV and Simian immunodeficiency virus (SIV) (93), as well as in cats coinfected with Feline immunodeficiency Virus (FIV) and FFV or FeLV and FFV (86; 94).

The seven miRNAs derived from BLV comprised no more than 8% of the total miRNA sequencing reads

in all analyzed samples (data not provided), consistent with the expression levels reported by Casas et

al. (2020) (85) for natural BLV infections. In contrast, Rosewick et al. (2013) (44) found that all ten

mature BLV-derived miRNAs accounted for about 40% of total miRNAs present in B-cell lymphomas

from sheep infected with BLV. Similarly, Ochiai et al. (2021) (95) noted that BLV-derived miRNAs made

up 38% of total miRNAs in Japanese black cattle diagnosed with enzootic bovine leukosis (B-cell

lymphoma). Additionally, Kincaid et al. (2012) (21) reported that BLV miRNAs were found in greater

quantities than cellular miRNAs in the BL3.1 cell line persistently infected with BLV. Although the relative

expression levels of individual BLV miRNAs differ across studies, blv-miR-B4-5p and blv-miR-B5-3p

appear to be either weakly expressed or undetectable. The dominant BLV-miRNA may change

depending on the disease stage or phase of infection analyzed. Overall, these findings suggest that BLV

miRNAs are highly expressed in *ex vivo* cell cultures, persistently infected cell lines, primary B-cell

tumors, and cattle affected by enzootic bovine leukosis. In cattle with subclinically BLV-infections, their

expression levels might vary based on several factors, including the infection route (natural or

experimental) and the animal's age (whether calf or adult cow), among other factors (21; 44; 83; 84; 85;

95). Conversely, the other BLV transcripts show little to no expression *in vivo*, highlighting the essential

function of BLV miRNAs in viral persistence (96; 97).

Likewise, three BFV-derived miRNAs (bfv-miR-BF1-5p, bfv-miR-BF1-3p, and bfv-miR-BF2-5p) have

been identified in MDBK cells that are persistently and recently infected with BFV, showing higher

expression levels during persistent infections. Furthermore, two BFV-miRNAs were found in the

peripheral blood leukocytes of cattle experimentally infected with BFV, utilizing a qPCR assay (23). This

research is the first to document the expression of these three BFV-derived miRNAs in natural setting

infections.

A previous study compared BLV seropositive and seronegative cattle but without testing for BLV

genomic material (85). The detection of BLV-miRNAs in seronegative cows suggests a previous

exposure to the virus, likely resulting in an inadequate immune response that did not generate a positive

result in the anti-BLV ELISA test. However, a longitudinal analysis of seroconversion was not performed.

This study found that all ten virus-derived miRNAs were consistently and highly expressed in the BLV(+)

group, whereas only a few showed weak expression in the BLV(-) group. Although undetected infections

cannot be entirely excluded, it is improbable that adult animals over two years old, which tested negative

via qPCR and ELISA at two intervals (three months apart), were unrecognized cases. The near-

complete alignment of BLV(-) reads to the bovine genome confirms their bovine origin, while a small

subset of reads mapped exclusively to viral miRNAs. Therefore, it suggests that the reads aligning to

BLV- and BFV-derived miRNAs in BLV(-) samples are likely bona fide and not the result of cross-

mapping or artifacts (data not shown). Moreover, sample mislabeling appears to be unlikely, if such an

issue had occurred, it would expect to observe expression profiles of viral miRNAs in BLV(+) samples

that mirrored those of BLV(-). However, our data did not show such patterns. The miRNA expression

profiles in BLV(-) samples were not consistent with those from BLV(+) samples, further arguing against

sample misidentification or cross-contamination. The potential for BLV- or BFV-miRNAs to function as

xeno-miRs and be transmitted from infected to uninfected animals remains a question for future

research.

Notably, the expression profiles of BLV(+) / BFV-miRNAs(+) and BLV(-) samples distinctly differentiate

the two groups of cows groups, as shown in the PCA plot. This indicates a shift in the host miRNA

expression profile between them.

Among the significant differentially expressed bovine miRNAs, bta-miR-375 was identified, which has

been previously reported to be associated with BLV infection in cattle (85; 95; 98), and suggested it as

an early biomarker for diagnosing enzootic bovine leukosis (EBL). Moreover, its levels effectively

distinguished EBL-affected cattle from asymptomatic cattle with high sensitivity and specificity (99).

In a similar manner, bta-miR-133a was found to be differentially expressed in the serum of BLV-

seropositive cows compared to seronegative cows (98). The other significantly differentially expressed

miRNAs (bta-miR-677, bta-miR-95, bta-miR-9-5p, bta-miR-3613a, and bta-miR-1) have not been

previously reported in relation to BLV and/or BFV infections.

To date, efforts to assign biological functions to BLV- and BFV-derived miRNAs have focused on

predicting potential target genes for specific BLV- and BFV-miRNAs, in addition to selecting candidate

target genes for functional assays (21; 41; 44; 45; 83; 84; 85). In this study, we evaluated the functional

interaction (protein-protein interaction network) of potential target genes of BLV-miRNAs, while also

considering target genes of BFV-miRNAs and differentially expressed bovine-miRNAs.

It has been demonstrated that BLV-miRNAs are crucial in promoting disease progression in cattle (83;

100). Additionally, BLV fails to induce leukemia/lymphoma in sheeps (oncogenicity suppressed)

challenged with an isogenic BLV provirus lacking the miRNA genomic region.

The functional annotation of the proteins within hub C1 revealed that immune system-related GO BP

terms, such as "*Cytokine-mediated signaling pathway*", "*Positive regulation of leukocyte chemotaxis*",

"*Regulation of leukocyte proliferation*", "*Positive regulation of T cell activation*", "*Regulation of leukocyte*

*apoptotic process*", "*Regulation of interferon-gamma production*", were significant overrepresented,

along with the KEGG and Reactome terms: "*Cytokine-cytokine receptor interaction*", "*Viral protein*

*interaction with cytokine and cytokine receptor*", "*Cytokine signaling in immune system*". The Figure 5 A)

showed the enrichment graph connecting gene target to BP GO terms. It is widely accepted that BLV

infection changes the cytokines expression patterns and alters how the immune system produces

cytokines in response to BLV antigen stimulation (101; 102). Particularly, BLV disease progression

would be related to the suppression of the cell-mediated immune response (103). For example,

Interleukin-10 (IL-10), a suppressor and anti-inflammatory immune response cytokine, is over expressed

in cows with persistent lymphocytosis (104), which can inhibit cytokine production by Th1 cells (for

example, IL-2, IL-12 and gamma interferon) (104; 105) influencing the B-cell proliferation and

differentiation (104; 106).

In turn, the analysis of overrepresented metabolic pathways for the complete potential target gene set

identified "*Pathway in cancer*" as one of the significant KEGG pathways. This pathway involves a series

of cellular signaling pathways that activate crucial hallmarks of tumorigenesis, including tissue invasion

and metastasis, evading apoptosis, genomic instability, cell proliferation, genomic damage, and

insensitivity to anti-growth signals (107). The overrepresented GO BP terms for the biological function

hub C2 included terms associated with cellular processes that, when disrupted, could be linked to tumor

development events. These terms included "*Transmembrane receptor protein tyrosine kinase signaling*

*pathway*", "*Vascular endothelial growth factor signaling pathway*", "*ERBB2-EGFR signaling pathway*",

"*Positive regulation of MAP kinase activity*", "*Hemopoiesis*", "*Negative regulation of apoptotic process*",

among others. Interestingly, other overrepresented KEGG terms in C2 included the "*MAPK signaling*

*pathway*", "*JAK-STAT signaling pathway*", "*HIF-1 signaling pathway*", "*Regulation of actin cytoskeleton*"

and Reactome terms such as "*PI3K/AKT Signaling*" and "*Signaling by Receptor Tyrosine Kinases*". This

signaling pathways are amplified and propagated intracellularly by various kinases, ultimately affecting

how transcription factors and histone-modifying complexes control downstream gene expression (108).

In this context, a crucial signaling cascade most altered in human cancers is the mitogen-activated

protein kinase (MAPK) pathway, which includes the RAS–RAF–MAPK kinase (MEK)–extracellular

signal-regulated kinase (ERK) pathway (109; 110). Using the ovine BLV pathogenesis model, sheeps

were experimentally challenged with an isogenic BLV-miRNA deletion mutant, and global transcriptome

analysis revealed that BLV-miRNAs primarily promote the proliferation of BLV-infected B-cells by

inhibiting immune response and cell signaling pathways (100). In turn, in a reporter assay, the FBJ

murine osteosarcoma viral oncogene homolog (*FOS*) was identified as a direct target of BLV-miRNAs.

The *FOS* gene is a component of the activator protein-1 (AP-1) complex, which is involved in the primary

response to B-cell receptor signaling and is frequently downregulated in many cancers, including breast

carcinomas (83; 111).

Additionally, enriched GO BP terms across C3 included "*Histone modification*", "*Chromatin remodeling*",

and "*Histone deacetylation*", "*Negative regulation of transcription by RNA polymerase II*", "*Regulation of*

*apoptotic signaling pathway*", among others; as well as the Reactome terms "*Regulation of TP53 Activity*

*through Acetylation*" and "*Chromatin modifying enzymes*". The transcription factor *TP53*, a tumor

suppressor activated by DNA damage, plays a critical role in maintaining genomic integrity (112). The

expression of two transcription factors, B-lymphocyte-induced maturation protein 1 (*BLIMP1*) and B-cell

lymphoma 6 (*BCL6*), is negatively correlated with BLV-miRNA expression in BLV(+) cows. These

transcription factors play a pivotal role regulating B-cell differentiation, antibody affinity and T-cells

immune function. Particularly, an important function of *BCL6* is to enable GC B-cells to proliferate in

response to T-cell antigens by specifically repressing gene expression related to DNA damage sensing.

This repression allows for the tolerance of DNA breaks induced during immunoglobulin class

rearrangement and somatic hypermutation (84; 113; 114).

The functional role of BFV-miRNAs remains poorly investigated. Predicted gene targets for bfv-miR-BF2-

5p were Ankyrin repeat domain-containing protein (*ANKRD17*) and Bax-interacting factor 1 (*BIF1*) (41).

*BIF1* plays a key role in activating the pro-apoptotic Bax protein in the intrinsic apoptosis pathway as well

as in autophagy and autophagosome formation, thus acting as a tumor suppressor (115; 116).

*ANKRD17,* on the other hand, is involved in DNA replication and cell cycle progression. It also interacts

with genes that are responsible for sensing viral RNAs and trigger immune responses (117; 118).

Notably, although target genes of bovine, BLV, and BFV miRNAs do not overlap, functional analysis

revealed convergence on similar cellular processes, implying coordinated regulatory interactions (Figure

5).

Regarding the differentially expressed BTA-miRNAs, miR-1, miR-133a, and miR-375 are recognized as

tumor suppressors and are downregulated in various types of cancer, exerting their effects through

distinct mechanisms (119; 120). For instance, miR-1 interacts with the proto-oncogene B-cell lymphoma

2 (*BCL-2*), an anti-apoptotic gene. Overexpression of miR-1 inhibits cell proliferation, migration, and

invasion while promoting apoptosis in breast cancer (121). Similarly, miR-133a modulates biological

processes such as proliferation, apoptosis, and autophagy (120). On the other hand, miR-375 is

significantly downregulated in several cancers and considered a biomarker of poor prognosis (122; 123;

124; 125), but it is overexpressed in breast cancer (126). Similarly, miR-95-3p and miR-9-3p act as

prognostic markers, promoting the progression of cervical and breast cancers, respectively (127; 128).

Conversely, the differentially expressed bovine miRNA bta-miR-677 enhances the production of type I

interferons (IFN-I) and interferon-stimulated genes (ISGs) (129). Similarly, inhibiting miR-3613-3p

decreases the expression of IFN-α and IFN-β, thereby affecting the anti-hepatitis B activity of interferons

(130). Moreover, miR-3613-3p has been recognized as a tumor suppressor, with its deletion associated

with poor prognosis in estrogen receptor-positive breast cancer patients (131).

**Conclusion**

In this study, we report for the first time the co-expression of seven miRNAs derived from BLV and three

from BFV in cattle naturally infected with BLV. Several bovine BTA-miRNAs were found to be

differentially expressed (DE) between cows with BLV high proviral load (PVL) and non-infected cows. Among them, bta-miR-375 was previously identified as a potential early biomarker for the progression of enzootic bovine leukosis, and this study further emphasizes its potential utility in identifying high PVL cows alongside other candidate miRNAs.

The functional analysis of protein-protein interaction networks involving BLV-, BFV-, and BTA-DE miRNAs-targeted genes identified key metabolic pathways that potentially underlie tumorigenesis and immune response modulation, including cellular signaling pathways, regulation of T cell activation, apoptosis, chromatin remodeling, and cytoskeleton regulation. Moreover, miRNA targets from both host and viral origins would act synergistically to influence cellular processes. The co-expression of miRNAs derived from BLV and BFV raises concerns about whether their interaction may exacerbate disease pathogenesis or facilitate the dissemination of BFV in light of the zoonotic potential that BFV presents. Ultimately, the identified miRNAs and associated metabolic pathways offer promising opportunities for developing diagnostic tools and therapeutic strategies aimed at controlling viral persistence and tumorigenesis in cattle.

## Declarations

## Competing interests

The authors declare that they have no competing interests.

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## Authors' contributions

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Data curation: HAC.

Formal analysis: MIP, HAC.

Funding acquisition: CM, JPJ, HAC

Investigation: MIP, GSA, CP, DDG, KGT, CM, JPJ, HAC.

Methodology: HAC.

Visualization: HAC.

Writing – original draft: HAC.

Writing – review & editing: MIP, GSA, CP, DDG, KGT, CM, JPJ, HAC.

Roles as defined by: CRediT (contributor role taxonomy)

## Availability of data

Raw small sequencing reads are available in the NCBI Sequence Read Archive (SRA) under BioProject accession number PRJNA1219522.

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## References

1. Ruivinho C, Gama-Carvalho M. 20 June 2023. Small non-coding RNAs encoded by RNA viruses: old controversies and new lessons from the COVID-19 pandemic. Front Genet. 21;14:1216890. https://doi.org/10.3389/fgene.2023.1216890.

2. Samols MA, Hu J, Skalsky RL, Renne R. 1 July 2005. Cloning and identification of a microRNA cluster within the latency-associated region of Kaposi's sarcoma-associated herpesvirus. J Virol. 79:9301-5. https://doi.org/10.1128/JVI.79.14.9301-9305.2005.

3. Sullivan CS, Ganem D. 7 October 2005. MicroRNAs and viral infection. Mol Cell.;20:3-7. https://doi.org/10.1016/j.molcel.2005.09.012.

4. Grundhoff A, Sullivan CS. 15 March 2011. Virus-encoded microRNAs. Virology. 15;411:325-43. https://doi.org/10.1016/j.virol.2011.01.002.

5. Guo YE, Steitz JA. 21 July 2014. Virus meets host microRNA: the destroyer, the booster, the hijacker. Mol Cell Biol. 34(20):3780-7. https://doi.org/10.1128/MCB.00871-14.

6. Bartel DP. 23 January 2004 MicroRNAs: genomics, biogenesis, mechanism, and function. Cell. 116(2):281-97. https://doi.org/10.1016/s0092-8674(04)00045-5.

7. Kim VN. 1 May 2005. MicroRNA biogenesis: coordinated cropping and dicing. Nat Rev Mol Cell Biol. 6(5):376-85. https://doi.org/10.1038/nrm1644.

8. Bartel DP. 23 January 2009. MicroRNAs: target recognition and regulatory functions. Cell. 136(2):215-33. https://doi.org/10.1016/j.cell.2009.01.002.

9. He L, Hannon GJ. 1 July 2004. MicroRNAs: small RNAs with a big role in gene regulation. Nat Rev Genet. 5(7):522-31. https://doi.org/10.1038/nrg1379.

10. Friedman RC, Farh KK, Burge CB, Bartel DP. January 2009. Most mammalian mRNAs are conserved targets
of microRNAs. Genome Res. 19(1):92-105. https://doi.org/10.1101/gr.082701.108.

11. Bartel DP. 22 March 2018. Metazoan MicroRNAs. Cell. 173(1):20-51. https://doi.org/10.1016/j.cell.2018.03.006.

12. Girardi E, López P, Pfeffer S. 1 October 2018 On the Importance of Host MicroRNAs During Viral Infection.
Front Genet. 9:439. https://doi.org/10.3389/fgene.2018.00439.

13. Skalsky RL, Cullen BR. 2010. Viruses, microRNAs, and host interactions. Annu Rev Microbiol. 64:123-41.
https://doi.org/10.1146/annurev.micro.112408.134243.

14. Pfeffer S, Zavolan M, Grässer FA, Chien M, Russo JJ, Ju J, John B, Enright AJ, Marks D, Sander C, Tuschl T.
30 April 2004. Identification of virus-encoded microRNAs. Science. 304(5671):734-6.
https://doi.org/10.1126/science.1096781.

15. Pfeffer S, Sewer A, Lagos-Quintana M, Sheridan R, Sander C, Grässer FA, van Dyk LF, Ho CK, Shuman S,
Chien M, Russo JJ, Ju J, Randall G, Lindenbach BD, Rice CM, Simon V, Ho DD, Zavolan M, Tuschl T. 16 February
2005. Identification of microRNAs of the herpesvirus family. Nat Methods. 2(4):269-76.
https://doi.org/10.1038/nmeth746.

16. Seo GJ, Fink LH, O'Hara B, Atwood WJ, Sullivan CS. 6 August 2008. Evolutionarily conserved function of a
viral microRNA. J Virol. 82(20):9823-8. https://doi.org/10.1128/JVI.01144-08.

17. Ghosh Z, Mallick B, Chakrabarti J. 18 December 2008. Cellular versus viral microRNAs in host-virus
interaction. Nucleic Acids Res. 37(4):1035-48. https://doi.org/10.1093/nar/gkn1004.

18. Hussain M, Torres S, Schnettler E, Funk A, Grundhoff A, Pijlman GP, Khromykh AA, Asgari S. 12 November
2011. West Nile virus encodes a microRNA-like small RNA in the 3' untranslated region which up-regulates GATA4
mRNA and facilitates virus replication in mosquito cells. Nucleic Acids Res. 40(5):2210-23.
https://doi.org/10.1093/nar/gkr848.

19. Hussain M, Asgari S. 3 February 2014 MicroRNA-like viral small RNA from Dengue virus 2 autoregulates its
replication in mosquito cells. Proc Natl Acad Sci US 111(7):2746-51. https://doi.org/ 10.1073/pnas.1320123111.

20. Liang H, Zhou Z, Zhang S, Zen K, Chen X, Zhang C. 29 September 2014. Identification of Ebola virus
microRNAs and their putative pathological function. Sci China Life Sci. 57(10):973-81.
https://doi.org/10.1007/s11427-014-4759-2.

21. Kincaid RP, Burke JM, Sullivan CS. 30 January 2012. RNA virus microRNA that mimics a B-cell oncomiR. Proc
Natl Acad Sci US 109(8):3077-82. https://doi.org/10.1073/pnas.1116107109.

22. Kincaid RP, Chen Y, Cox JE, Rethwilm A, Sullivan CS. 8 April 2014. Noncanonical microRNA (miRNA)
biogenesis gives rise to retroviral mimics of lymphoproliferative and immunosuppressive host miRNAs. mBio
5(2):e00074. https://doi.org/10.1128/mBio.00074-14.

23. Whisnant AW, Kehl T, Bao Q, Materniak M, Kuzmak J, Löchelt M, Cullen BR. 1 May 2014. Identification of
novel, highly expressed retroviral microRNAs in cells infected by bovine foamy virus. J Virol. 88(9):4679-86.
https://doi.org/10.1128/JVI.03587-13.

24. Yao Y, Smith LP, Nair V, Watson M. January 2014. An avian retrovirus uses canonical expression and
processing mechanisms to generate viral microRNA. J Virol. 88(1):2-9. https://doi.org/10.1128/JVI.02921-13.

25. Teng Y, Wang Y, Zhang X, Liu W, Fan H, Yao H, Lin B, Zhu P, Yuan W, Tong Y, Cao W. 26 May 2015.
Systematic Genome-wide Screening and Prediction of microRNAs in EBOV During the 2014 Ebolavirus Outbreak.
Sci Rep. 5:9912. https://doi.org/10.1038/srep09912.

26. Cao W, Heit A, Hotz-Wagenblatt A, Löchelt M. 31 May 2018. Functional characterization of the bovine foamy
virus miRNA expression cassette and its dumbbell-shaped pri-miRNA. Virus Genes. 54(4):550-560.
https://doi.org/10.1007/s11262-018-1574-z.

27. Hashimoto-Gotoh A, Kitao K, Miyazawa T. 23 January 2020. Persistent Infection of Simian Foamy Virus
Derived from the Japanese Macaque Leads to the High-Level Expression of microRNA that Resembles the miR-1
microRNA Precursor Family. Microbes Environ. 35(1):ME19130. https://doi.org/10.1264/jsme2.ME19130.

28. Aso S, Kitao K, Hashimoto-Gotoh A, Sakaguchi S, Miyazawa T. 13 November 2021. Identification of Feline
Foamy Virus-derived MicroRNAs. Microbes Environ. 36(4):ME21055. https://doi.org/10.1264/jsme2.ME21055.

29. Harwig A, Das AT, Berkhout B. 23 April 2014. Retroviral microRNAs. Curr Opin Virol. 7:47-54.
https://doi.org/10.1016/j.coviro.2014.03.013.

30. Aguado LC, tenOever B. 31 May 2018. RNA virus building blocks-miRNAs not included. PLoS Pathog.
14(5):e1006963. https://doi.org/10.1371/journal.ppat.1006963.

31. Vasudevan S, Tong Y, Steitz J. 21 December 2007. Switching from Repression to Activation: MicroRNAs Can
Up-Regulate Translation. Science. 318, 5858:1931-1934. https://doi.org/10.1126/science.1149460.
32. Chenarboo F, Ng H, Macdonald D, Mahal L. 23 November 2022. High-Throughput Analysis Reveals miRNA
Upregulating α-2,6-Sialic Acid through Direct miRNA–mRNA Interactions. ACS Cent. Sci. 8, 11, 1527–1536.
https://doi.org/10.1021/acscentsci.2c00748.
33. Cullen BR. 22 January 2009. Viral and cellular messenger RNA targets of viral microRNAs. Nature.
22;457(7228):421-5. https://doi.org/10.1038/nature07757.
34. Cullen BR. 15 February 2013. MicroRNAs as mediators of viral evasion of the immune system. Nat Immunol.
14(3):205-10. https://doi.org/10.1038/ni.2537.
35. Tycowski KT, Guo YE, Lee N, Moss WN, Vallery TK, Xie M, Steitz JA. 15 March 2015. Viral noncoding RNAs:
more surprises. Genes Dev. 29(6):567-84. https://doi.org/10.1101/gad.259077.115.
36. Pluta A, Jaworski JP, Douville RN. 25 September 2020. Regulation of Expression and Latency in BLV and
HTLV. Viruses. 12(10):1079. https://doi.org/10.3390/v12101079.
37. Glazov EA, Horwood PF, Assavalapsakul W, Kongsuwan K, Mitchell RW, Mitter N, Mahony TJ. 1 January
2010. Characterization of microRNAs encoded by the bovine herpesvirus 1 genome. J Gen Virol. 91:32-41.
https://doi.org/10.1099/vir.0.014290-0.
38. Tang Q, Wu YQ, Chen DS, Zhou Q, Chen HC, Liu ZF. 1 March 2014. Bovine herpesvirus 5 encodes a unique
pattern of microRNAs compared with bovine herpesvirus 1. J Gen Virol. 95:671-678.
https://doi.org/10.1099/vir.0.061093-0.
39. Jaber T, Workman A, Jones C. 21 April 2010. Small noncoding RNAs encoded within the bovine herpesvirus 1
latency-related gene can reduce steady-state levels of infected cell protein 0 (bICP0). J Virol. 84(13):6297-307.
https://doi.org/10.1128/JVI.02639-09.
40. da Silva LF, Jones C. February 2012. Two microRNAs encoded within the bovine herpesvirus 1 latency-related
gene promote cell survival by interacting with RIG-I and stimulating NF-κB-dependent transcription and beta
interferon signaling pathways. J Virol. 86(3):1670-82. https://doi.org/10.1128/JVI.06550-11.
41. Cao W, Stricker E, Hotz-Wagenblatt A, Heit-Mondrzyk A, Pougialis G, Hugo A, Kuźmak J, Materniak-Kornas M,
Löchelt M. 2 November 2020. Functional Analyses of Bovine Foamy Virus-Encoded miRNAs Reveal the
Importance of a Defined miRNA for Virus Replication and Host-Virus Interaction. Viruses. 12(11):1250.
https://doi.org/10.3390/v12111250.
42. Han YC, Park CY, Bhagat G, Zhang J, Wang Y, Fan JB, et al. 15 March 2010. microRNA-29a induces aberrant
self-renewal capacity in hematopoietic progenitors, biased myeloid development, and acute myeloid leukemia. J
Exp Med. 207: 475–489. https://doi.org/10.1084/jem.20090831.
43 Santanam U, Zanesi N, Efanov A, Costinean S, Palamarchuk A, Hagan JP, et al. 21 June 2010 Chronic
lymphocytic leukemia modeled in mouse by targeted miR-29 expression. Proc Natl Acad Sci U S A. 107: 12210–
12215. https://doi.org/10.1073/pnas.1007186107.
44. Rosewick N, Momont M, Durkin K, Takeda H, Caiment F, Cleuter Y, Vernin C, Mortreux F, Wattel E, Burny A,
Georges M, Van den Broeke A. 23 january 2013. Deep sequencing reveals abundant noncanonical retroviral
microRNAs in B-cell leukemia/lymphoma. Proc Natl Acad Sci U S A. 110(6):2306-11.
https://doi.org/10.1073/pnas.1213842110.
45. Petersen MI, Carignano HA, Mongini C, Gonzalez DD, Jaworski JP. 2 February 2023. Bovine leukemia virus
encoded blv-miR-b4-3p microRNA is associated with reduced expression of anti-oncogenic gene in vivo. PLoS
One 18(2):e0281317. https://doi.org/10.1371/journal.pone.0281317.
46. VanLeeuwen JA, Forsythe L, Tiwari A, Chartier R. January 2005. Seroprevalence of antibodies against bovine
leukemia virus, bovine viral diarrhea virus, Mycobacterium avium subspecies paratuberculosis, and Neospora
caninum in dairy cattle in Saskatchewan. Can Vet J. 46:56–58.
47. VanLeeuwen JA, Tiwari A, Plaizier JC, Whiting TL. August 2006. Seroprevalences of antibodies against bovine
leukemia virus, bovine viral diarrhea virus, Mycobacterium avium subspecies paratuberculosis, and Neospora
caninum in beef and dairy cattle in Manitoba. Can Vet J. 47:783–786.
48. NAHMS. 2007.
https://www.aphis.usda.gov/animal_health/nahms/dairy/downloads/dairy07/Dairy07_is_BLV.pdf.

49. Gutiérrez G, Carignano H, Alvarez I, Martínez C, Porta N, Politzki R, Gammella M, Lomonaco M, Fondevila N,
Poli M, Trono K. 9 October 2012. Bovine leukemia virus p24 antibodies reflect blood proviral load. BMC Vet Res.
8:187. https://doi.org/10.1186/1746-6148-8-187.
50. OIE. Manual of Diagnostic Tests and Vaccines for Terrestrial Animals. 2012. 7th Ed.
51. Erskine RJ, Bartlett PC, Byrem TM, Render CL, Febvay C, Houseman JT. 27 March 2012. Using a herd profile
to determine age-specific prevalence of bovine leukemia virus in Michigan dairy herds. Vet Med Int.
https://doi.org/10.1155/2012/350374.
52. Bartlett PC, Norby B, Byrem TM, Parmelee A, Ledergerber JT, Erskine RJ. March 2013. Bovine leukemia virus
and cow longevity in Michigan dairy herds. J Dairy Sci. 96:1591–1597. https://doi.org/10.3168/jds.2012-5930.
53. Materniak-Kornas M, Tan J, Heit-Mondrzyk A, Hotz-Wagenblatt A, Löchelt M. 21 November 2019. Bovine
Foamy Virus: Shared and Unique Molecular Features In Vitro and In Vivo. Viruses. 11(12):1084.
https://doi.org/10.3390/v11121084.
54. De Brun L, Mionetto M, Rodríguez F, Brandl S, Hubner S, Fischer G, Puentes R. 15 July 2023. Natural
Association between Bovine Leukemia Virus and Reproductive Infectious Diseases. Acta Scientiae Veterinariae.
51:1922. https://doi.org/10.22456/1679-9216.129251
55. Méndez A, Cerón-Téllez F, Sarmiento R, Tórtora-Pérez J, Rojas-Anaya E, Álvarez H. April 2023. Presence of
co-infection between bovine leukemia virus and bovine herpesvirus 1 in herds vaccinated against bovine
respiratory complex. Canadian Journal of Veterinary Research. 87(2):105-109(5).
56. Oviedo-Pastrana M, Doria-Ramos M, Mattar S, Oviedo-Socarras T, Vallejo-Timarán D. 4 August 2024.
Seroprevalence of bovine leukemia virus and association with bovine infectious abortion in Creole breeds from
tropical grazing herds in the Colombian Caribbean. Vet World. 17(8):1715-1721.
https://doi.org/10.14202/vetworld.2024.1715-1721.
57. Petersen MI, Carignano HA, Suarez Archilla G, Caffaro ME, Alvarez I, Miretti MM, Trono K. February 2021.
Expression-based analysis of genes related to single nucleotide polymorphism hits associated with bovine
leukemia virus proviral load in Argentinean dairy cattle. J Dairy Sci. 104(2):1993-2007.
https://doi.org/10.3168/jds.2020-18924.
58. Alvarez I, Gutiérrez G, Gammella M, Martínez C, Politzki R, González C, Caviglia L, Carignano H, Fondevila N,
Poli M, Trono K. May 2013. Evaluation of total white blood cell count as a marker for proviral load of bovine
leukemia virus in dairy cattle from herds with a high seroprevalence of antibodies against bovine leukemia virus.
Am J Vet Res.;74(5):744-9. https://doi.org/10.2460/ajvr.74.5.744.
59. Trono KG, Pérez-Filgueira DM, Duffy S, Borca MV, Carrillo C. 26 November 2001. Seroprevalence of bovine
leukemia virus in dairy cattle in Argentina: comparison of sensitivity and specificity of different detection methods.
Vet Microbiol. 83(3):235-48. https://doi.org/10.1016/s0378-1135(01)00420.
60. Petersen MI, Alvarez I, Trono KG, Jaworski JP. July 2018. Quantification of bovine leukemia virus proviral DNA
using a low-cost real-time polymerase chain reaction. J Dairy Sci. 101(7):6366-6374.
https://doi.org/10.3168/jds.2017-14253.
61. Hopkins SG, DiGiacomo RF. March 1997. Natural transmission of bovine leukemia virus in dairy and beef
cattle. Vet Clin North Am Food Anim Pract. 13(1):107-28. https://doi.org/10.1016/s0749-0720(15)30367-4.
62. Smith T, Heger A, Sudbery I. March 2017. UMI-tools: modeling sequencing errors in Unique Molecular
Identifiers to improve quantification accuracy. Genome Res. 27(3):491-499. https://doi.org/10.1101/gr.209601.116.
63. Martin, M. 2011. Cutadapt removes adapter sequences from high-throughput sequencing reads.
EMBnet.journal. 17(1):pp10-12. https://doi.org/10.14806/ej.17.1.200.
64. Bushnell, B. 21 March 2014. BBMap: A Fast, Accurate, Splice-Aware Aligner. Berkeley: Lawrence Berkeley
National Lab. Report #: LBNL-7065E. https://escholarship.org/uc/item/1h3515gn.
65. Langmead B, Salzberg SL. 4 March 2012. Fast gapped-read alignment with Bowtie 2. Nat Methods. 9(4):357-9.
https://doi.org/10.1038/nmeth.1923.
66. Liu D. 16 December 2019. Algorithms for efficiently collapsing reads with Unique Molecular Identifiers. PeerJ.
7:e8275. https://doi.org/10.7717/peerj.8275.
67. Kozomara A, Birgaoanu M, Griffiths-Jones S. 8 January 2019. miRBase: from microRNA sequences to
function. Nucleic Acids Res. 47(D1):D155-D162. https://doi.org/10.1093/nar/gky1141.

68. Friedländer M, Mackowiak S, Li N, Chen W, Rajewsky N. January 2012. miRDeep2 identifica con precisión genes de microARN conocidos y cientos de nuevos en siete clados animales, Nucleic Acids Research 40(1):37–52. https://doi.org/10.1093/nar/gkr688.

69. Love MI, Huber W, Anders S. 5 December 2014. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. Genome Biol. 15(12):550. https://doi.org/10.1186/s13059-014-0550-8.

70. Benjamini Y, Hochberg Y. 1995. Controlling the false discovery rate: A practical and powerful approach to multiple testing. J Roy Statist Soc Ser B. 57:289–300. https://doi.org/10.1111/j.2517-6161.1995.tb02031.x.

71. R Core Team, 2024. R: A Language and Environment for Statistical Computing. Vienna, Austria.

72. Enright AJ, John B, Gaul U, Tuschl T, Sander C, Marks DS. 12 December 2003. MicroRNA targets in Drosophila. Genome Biol. 5(1):R1. https://doi.org/10.1186/gb-2003-5-1-r1.

73. Kertesz M, Iovino N, Unnerstall U, Gaul U, Segal E. 23 September 2007. The role of site accessibility in microrna target recognition. Nature genetics, 39(10):1278–1284. https://doi.org/10.1038/ng2135.

74. Krüger J, Rehmsmeier M. 1 July 2006. RNAhybrid: microRNA target prediction easy, fast and flexible. Nucleic Acids Res, 34, p.W451-4. https://doi.org/10.1093/nar/gkl243.

75. Rosen BD, Bickhart DM, Schnabel RD, Koren S, Elsik CG, Tseng E, Rowan TN, Low WY, Zimin A, Couldrey C, Hall R, Li W, Rhie A, Ghurye J, McKay SD, Thibaud-Nissen F, Hoffman J, Murdoch BM, Snelling WM, McDaneld TG, Hammond JA, Schwartz JC, Nandolo W, Hagen DE, Dreischer C, Schultheiss SJ, Schroeder SG, Phillippy AM, Cole JB, Van Tassell CP, Liu G, Smith TPL, Medrano JF. 1 March 2020. De novo assembly of the cattle reference genome with single-molecule sequencing. Gigascience. 1;9(3):giaa021. https://doi.org/10.1093/gigascience/giaa021.

76. Szklarczyk D, Gable AL, Nastou KC, Lyon D, Kirsch R, Pyysalo S, Doncheva NT, Legeay M, Fang T, Bork P, Jensen LJ, von Mering C. 8 January 2021. The STRING database in 2021: customizable protein-protein networks, and functional characterization of user-uploaded gene/measurement sets. Nucleic Acids Res. 49(D1):D605-D612. doi: 10.1093/nar/gkaa1074. Erratum in: Nucleic Acids Res. 2021 Oct 11;49(18):10800. https://doi.org/10.1093/nar/gkab835.

77. Kanehisa M, Goto S. 1 January 2000. KEGG: kyoto encyclopedia of genes and genomes. Nucleic Acids Res. 28(1):27-30. https://doi.org/10.1093/nar/28.1.27.

78. Tavazoie S, Hughes JD, Campbell MJ, Cho RJ, Church GM. July 1999. Systematic determination of genetic network architecture. Nat Genet. 22(3):281-5. https://doi.org/10.1038/10343.

79. Sayols S. 18 April 2023. rrvgo: a Bioconductor package for interpreting lists of Gene Ontology terms. MicroPubl Biol. https://doi.org/10.17912/micropub.biology.000811.

80. Mirsky ML, Olmstead CA, Da Y, Lewin HA. 1996. The prevalence of proviral bovine leukemia virus in peripheral blood mononuclear cells at two subclinical stages of infection. J Virol. 70:2178–2183. https://doi.org/10.1128/jvi.70.4.2178-2183.1996.

81. Jimba M, Takeshima S, Matoba K, Endoh D, Aida Y. 2 November 2010. BLV-CoCoMo-qPCR: quantitation of bovine leukemia virus proviral load using the CoCoMo algorithm. Retrovirology. 7:91. https://doi.org/10.1186/1742-4690-7-91.

82. Mammerickx M, Portetelle D, de Clercq K, Burny A. 1987. Experimental transmission of enzootic bovine leukosis to cattle, sheep and goats: infectious doses of blood and incubation period of the disease. Leuk Res.11:353–358. https://doi.org/10.1016/0145-2126(87)90180-9.

83. Gillet NA, Hamaidia M, de Brogniez A, Gutiérrez G, Renotte N, Reichert M, Trono K, Willems L. 28 April 2016. Bovine Leukemia Virus Small Noncoding RNAs Are Functional Elements That Regulate Replication and Contribute to Oncogenesis In Vivo. PLoS Pathog. 12(4):e1005588. https://doi.org/10.1371/journal.ppat.1005588.

84. Frie MC, Droscha CJ, Greenlick AE, Coussens PM. 15 January 2018. MicroRNAs Encoded by Bovine Leukemia Virus (BLV) Are Associated with Reduced Expression of B Cell Transcriptional Regulators in Dairy Cattle Naturally Infected with BLV. Front Vet Sci. 4:245. https://doi.org/10.3389/fvets.2017.00245.

85. Casas E, Ma H, Lippolis JD. 21 September 2020. Expression of Viral microRNAs in Serum and White Blood Cells of Cows Exposed to Bovine Leukemia Virus. Front Vet Sci. 7:536390. https://doi.org/10.3389/fvets.2020.536390.

86. Cavalcante LTF, Muniz CP, Jia H, Augusto AM, Troccoli F, Medeiros SO, Dias CGA, Switzer WM, Soares MA, Santos AF. 11 December 2018. Clinical and Molecular Features of Feline Foamy Virus and Feline Leukemia Virus Co-Infection in Naturally-Infected Cats. Viruses. 10(12):702. https://doi.org/10.3390/v10120702.

87. Alais S, Pasquier A, Jegado B, Journo C, Rua R, Gessain A, Tobaly-Tapiero J, Lacoste R, Turpin J, Mahieux
R. 1 October 2018.  STLV-1 co-infection is correlated with an increased SFV proviral load in the peripheral blood of
SFV/STLV-1 naturally infected non-human primates. PLoS Negl Trop Dis. 12(10):e0006812.
https://doi.org/10.1371/journal.pntd.0006812.
88. Amborski GF, Lo L, Seger CL. July 1989. Serological detection of multiple retroviral infections in cattle: Bovine
leukemia virus, bovine syncytial virus and bovine visna virus. Vet. Microbiol. 20:247–253.
https://doi.org/10.1016/0378-1135(89)90048-5.
89. Jacobs RM, Smith HE, Gregory B, Valli VE, Whetstone CA. October 1992. Detection of multiple retroviral
infections in cattle and cross-reactivity of bovine immunodeficiency-like virus and human immunodeficiency virus
type 1 proteins using bovine and human sera in a western blot assay. Can. J. Vet. Res. 56:353–359.
90. Materniak M., Hechler T., Löchelt M., Kuzmak J. March 2013. Similar patterns of infection with bovine foamy
virus in experimentally inoculated calves and sheep. J. Virol. 87:3516–3525. https://doi.org/10.1128/JVI.02447-12
91. Romen F, Backes P, Materniak M, Sting R, Vahlenkamp TW, Riebe R, Pawlita M, Kuzmak J, Löchelt M. 20 July
2007. Serological detection systems for identification of cows shedding bovine foamy virus via milk. Virology
364:123–131. https://doi.org/10.1016/j.virol.2007.03.009.
92. Pinto-Santini DM, Stenbak CR, Linial ML. 2 December 2017. Foamy virus zoonotic infections. Retrovirology.
14:55. doi: 10.1186/s12977-017-0379-9. https://doi.org/10.1186/s12977-017-0379-9.
93. Choudhary A, Galvin TA, Williams DK, Beren J, Bryant MA, Khan AS. 6 June 2013 Influence of naturally
occurring simian foamy viruses (SFVs) on SIV disease progression in the Rhesus macaque (Macaca mulatta)
model. Viruses. 5(6):1414–1430. https://doi.org/10.3390/v5061414.
94. Powers JA, Chiu ES, Kraberger SJ, Roelke-Parker M, Lowery I, Erbeck K, Troyer R, Carver S, VandeWoude S.
29 August 2018. Feline leukemia virus (FELV) disease outcomes in a domestic cat breeding colony: Relationship
to endogenous felv and other chronic viral infections. J. Virol. 92 https://doi.org/10.1128/JVI.00649-18.
95. Ochiai C, Miyauchi S, Kudo Y, Naruke Y, Yoneyama S, Tomita K, Dongze L, Chiba Y, Hirata TI, Ichijo T, Nagai
K, Kobayashi S, Yamada S, Hikono H, Murakami K. 2021 Characterization of microRNA expression in B cells
derived from Japanese black cattle naturally infected with bovine leukemia virus by deep sequencing. PLoS One.
16(9):e0256588. https://doi.org/10.1371/journal.pone.0256588.
96. Lagarias DM, Radke K. May 1989. Transcriptional activation of bovine leukemia virus in blood cells from
experimentally infected, asymptomatic sheep with latent infections. J Virol. 63(5):2099-107.
https://doi.org/10.1128/JVI.63.5.2099-2107.1989.
97. Radke K, Sigala T, Grossman D. May 1992. Transcription of bovine leukemia virus in peripheral blood cells
obtained during early infection in vivo. Microbial Pathogenesis 12:319-331 https://doi.org/10.1016/0882-
4010(92)90095-6.
98. Ma H, Lippolis JD, Casas E. 18 July 2022. Expression Profiles and Interaction of MicroRNA and Transcripts in
Response to Bovine Leukemia Virus Exposure. Front Vet Sci. 9:887560. https://doi.org/10.3389/fvets.2022.887560.
99. Murakami K, Matsunaga T, Matsuzaki T, Naruke Y, Miyauchi S, Kobayashi S, Yoneyama S, Sakai Y, Ichijo T,
Hirata TI, Kimura A, Chiba Y, Matsuda KI, Yamada S, Hikono H. 2024. Serum bta-miRNA-375 as a potential
biomarker for the early diagnosis of enzootic bovine leukosis. PLoS One. 19(5):e0302868.
https://doi.org/10.1371/journal.pone.0302868.
100. Safari R, Jacques JR, Brostaux Y, Willems L. 14 May 2020. Ablation of non-coding RNAs affects bovine
leukemia virus B lymphocyte proliferation and abrogates oncogenesis. PLoS Pathog. 16(5):e1008502.
https://doi.org/10.1371/journal.ppat.1008502.
101. Gillet N, Florins A, Boxus M, Burteau C, Nigro A, Vandermeers F, Balon H, Bouzar AB, Defoiche J, Burny A,
Reichert M, Kettmann R, Willems L. 16 March 2007. Mechanisms of leukemogenesis induced by bovine leukemia
virus: prospects for novel anti-retroviral therapies in human. Retrovirology. 4:18. https://doi.org/10.1186/1742-4690-
4-18.
102. Frie MC, Coussens PM. Bovine leukemia virus: A major silent threat to proper immune responses in cattle. 15
February 2015. Vet. Imm. and Imm. 163(3-4):103-114. https://doi.org/10.1016/j.vetimm.2014.11.014.
103. Kabeya H, Ohashi K, Onuma M. July 2001. Host immune responses in the course of bovine leukemia virus
infection. J Vet Med Sci. 63: 703-708. https://doi.org/10.1292/jvms.63.703.

104. Pyeon D, O'Reilly KL, Splitter GA. August 1996. Increased interleukin-10 mRNA expression in tumor-bearing
or persistently lymphocytotic animals infected with bovine leukemia virus. J Virol. 70(8):5706-10.
https://doi.org/10.1128/JVI.70.8.5706-5710.1996.
105. Amills M, Ramiya V, Norimine J, Olmstead CA, Lewin HA. 5 December 2002. Reduced IL-2 and IL-4 mRNA
expression in CD4+ T cells from bovine leukemia virus-infected cows with persistent lymphocytosis. Virology. 304:
1-9. https://doi.org/10.1006/viro.2002.1651.
106. Yakobson B, Brenner J, Ungar-Waron H, Trainin Z. 31 July 1998. Short-termed expression of interleukin-12
during experimental BLV infection may direct disease progression to persistent lymphocytosis. Vet Immunol
Immunopathol. 64: 207-218. https://doi.org/10.1016/S0165-2427(98)00136-6.
107. Weinberg, R.A. 2007. The Biology of Cancer Garland Science. Taylor & Francis Group, LLC, New York, NY.
108. Suganuma T, Workman JL. June 2013. Chromatin and signaling. Curr Opin Cell Biol. 25(3):322-6.
https://doi.org/10.1016/j.ceb.2013.02.016.
109. Gimple RC, Wang X. 24 September 2019. RAS: Striking at the Core of the Oncogenic Circuitry. Front Oncol.
9:965. https://doi.org/10.3389/fonc.2019.00965.
110. Bahar ME, Kim HJ, Kim DR. 18 December 2023 Targeting the RAS/RAF/MAPK pathway for cancer therapy:
from mechanism to clinical studies. Signal Transduct Target Ther. 8(1):455. https://doi.org/10.1038/s41392-023-
01705-z.
111. Huo L, Rothstein TL. 1 April 1995. Receptor-specific induction of individual AP-1 components in B
lymphocytes. J Immunol. 154(7):3300-9.
112. Aubrey BJ, Strasser A, Kelly GL. 2 May 2016. Tumor-Suppressor Functions of the TP53 Pathway. Cold
Spring Harb Perspect Med. 6(5):a026062. https://doi.org/10.1101/cshperspect.a026062.
113. Nutt SL, Fairfax KA, Kallies A. December 2007. BLIMP1 guides the fate of effector B and T cells. Nat Rev
Immunol. 7(12):923-7. https://doi.org/10.1038/nri2204.
114. Swaminathan S, Duy C, Müschen M. March 2014. BACH2-BCL6 balance regulates selection at the pre-B cell
receptor checkpoint. Trends Immunol. 35(3):131-7. https://doi.org/10.1016/j.it.2013.11.002.
115. Coppola D, Khalil F, Eschrich SA, Boulware D, Yeatman T, Wang HG. 15 November 2008. Down-regulation of
Bax-interacting factor-1 in colorectal adenocarcinoma. Cancer. 113(10):2665-70.
https://doi.org/10.1002/cncr.23892.
116. Xu L, Wang Z, He SY, Zhang SF, Luo HJ, Zhou K, Li XF, Qiu SP, Cao KY. December 2016. Bax-interacting
factor-1 inhibits cell proliferation and promotes apoptosis in prostate cancer cells. Oncol Rep. 36(6):3513-3521.
https://doi.org/10.3892/or.2016.5172.
117. Deng M, Li F, Ballif BA, Li S, Chen X, Guo L, Ye X. 20 March 2009. Identification and functional analysis of a
novel cyclin e/cdk2 substrate ankrd17. J Biol Chem. 284(12):7875-88. https://doi.org/10.1074/jbc.M807827200.
118. Wang Y, Tong X, Li G, Li J, Deng M, Ye X. May 2012. Ankrd17 positively regulates RIG-I-like receptor (RLR)-
mediated immune signaling. Eur J Immunol. 42(5):1304-15. https://doi.org/10.1002/eji.201142125.
119. Yoshino H, Chiyomaru T, Enokida H, Kawakami K, Tatarano S, Nishiyama K, Nohata N, Seki N, Nakagawa
M. 1 March 2011. The tumour-suppressive function of miR-1 and miR-133a targeting TAGLN2 in bladder cancer.
Br J Cancer. 104(5):808-18. https://doi.org/10.1038/bjc.2011.23.
120. Hua YT, Xu WX, Li H, Xia M. 1 January 2021. Emerging roles of MiR-133a in human cancers. J Cancer.
12(1):198-206. https://doi.org/10.7150/jca.48769.
121. Peng J, Yuan C, Wu Z, Wang Y, Yin W, Lin Y, Zhou L, Lu J. 4 May 2020. Upregulation of microRNA 1 inhibits
proliferation and metastasis of breast cancer. Mol Med Rep. 22(1):454-464.
https://doi.org/10.3892/mmr.2020.11111.
122. Mathé EA, Nguyen GH, Bowman ED, Zhao Y, Budhu A, Schetter AJ, Braun R, Reimers M, Kumamoto K,
Hughes D, Altorki NK, Casson AG, Liu CG, Wang XW, Yanaihara N, Hagiwara N, Dannenberg AJ, Miyashita M,
Croce CM, Harris CC. 2009. MicroRNA expression in squamous cell carcinoma and adenocarcinoma of the
esophagus: associations with survival. Clin Cancer Res. 15(19):6192-200. https://doi.org/10.1158/1078-0432.CCR-
09-1467.
123. Avissar M, Christensen BC, Kelsey KT, Marsit CJ. 15 April 2009. MicroRNA expression ratio is predictive of
head and neck squamous cell carcinoma. Clin Cancer Res. 15(8):2850-5. https://doi.org/10.1158/1078-0432.CCR-
08-3131.

124. Yan J, She Q, Shen X, Zhang Y, Liu B, Zhang G. 13 November 2017. Potential Role of MicroRNA-375 as Biomarker in Human Cancers Detection: A Meta-Analysis. Biomed Res Int. 2017:1875843. https://doi.org/10.1155/2017/1875843.

125. He Y, Jin J, Wang L, Hu Y, Liang D, Yang H, Liu Y, Shan B. 2017. Evaluation of miR-21 and miR-375 as prognostic biomarkers in oesophageal cancer in high-risk areas in China. Clin Exp Metastasis. 34(1):73-84. https://doi.org/10.1007/s10585-016-9828-4.

126. de Souza Rocha Simonini P, Breiling A, Gupta N, Malekpour M, Youns M, Omranipour R, Malekpour F, Volinia S, Croce CM, Najmabadi H, Diederichs S, Sahin O, Mayer D, Lyko F, Hoheisel JD, Riazalhosseini Y. 15 November 2010. Epigenetically deregulated microRNA-375 is involved in a positive feedback loop with estrogen receptor alpha in breast cancer cells. Cancer Res. 70(22):9175-84. https://doi.org/10.1158/0008-5472.CAN-10-1318.

127. Li X, Zeng Z, Wang J, Wu Y, Chen W, Zheng L, Xi T, Wang A, Lu Y. February 2020. MicroRNA-9 and breast cancer. Biomed Pharmacother. 122:109687. https://doi.org/10.1016/j.biopha.2019.109687.

128. Feng S, Lu Y, Sun L, Hao S, Liu Z, Yang F, Zhang L, Wang T, Jiang L, Zhang J, Liu S, Pang H, Wang Z, Wang H. 15 November 2022. MiR-95-3p acts as a prognostic marker and promotes cervical cancer progression by targeting VCAM1. Ann Transl Med. 10(21):1171. https://doi.org/10.21037/atm-22-5184.

129. Zhong C, Liao Z, Zhang B, Xiao L, Li J, Zhu X. August 2022. Bta-miR-677 contribute to interferon pathway affecting the proliferation of caprine parainfluenza virus type 3. Microb Pathog. 169:105642. https://doi.org/10.1016/j.micpath.2022.1056.

130. Zhao Y, Yu Y, Ye L. October 2019. MiR-3613-3p impairs IFN-induced immune response by targeting CMPK1 in chronic hepatitis B. Infect Genet Evol. 74:103919. https://doi.org/10.1016/j.meegid.2019.103919.

131. Chen C, Pan Y, Bai L, Chen H, Duan Z, Si Q, Zhu R, Chuang TH, Luo Y. 25 January 2021. MicroRNA-3613-3p functions as a tumor suppressor and represents a novel therapeutic target in breast cancer. Breast Cancer Res. 23(1):12. https://doi.org/10.1186/s13058-021-01389-9.

**Figure Legends**

Figure 1. Principal component analysis of miRNAs expression matrix. Blue triangles and Red circles represent BLV(+) and BLV(-) samples, respectively. PC = principal component. PC1 and PC2, represent 62% and 17% of variance contained in the expression matrix data, respectively.

Figure 2. Volcano plot summarizing the results of the differential expression analysis of bovine miRNAs between BLV(+) and BLV(-) groups. Each circle represents a bovine miRNA. Red circles indicate miRNAs with FC < |1.5| and non-significant, while blue circles represent miRNAs with FC > |1.5| and q < 0.05 (-log(q-value) > 1.3). Viral-derived miRNAs, primarily quantified in BLV(+) samples, were excluded from the plot. Vertical dashed lines represent a |1.5| FC in miRNA expression. Horizontal dashed line represents the significance cutoff q-value = 0.05.

Figure 3. Word cloud for GO term overrepresented in target genes for Biological Process A) and Molecular Fuction B) categories.

Figure 4. Protein-protein interaction (PPI) cluster analysis. Three cluster (> 15 nodes) were identified A) Cluster 1 (C1): Nº nodes: 21 - Nº edges: 81, avg. node degree: 7.7; B) Cluster 2 (C2): Nº nodes: 20 – Nº edges: 64, avg. node degree: 6.4; C) Cluster 3 (C3): Nº nodes: 16 – Nº edges: 34, avg. node degree: 4.2. Green nodes: predicted BTA-miRNA-DE protein target; Blue nodes: predicted BFV-miRNA protein target; Red nodes: predicted BLV-miRNA protein target. Edges represent protein-protein interactions as determined by STRINGdb evidence, including: known interactions (curated databases and/or experimentally validated), predicted interactions (gene fusions, gene co-occurrence), text mining, co-expression, and protein homology.

947 Figure 5. Enrichment graph plot of PPI clusters. Each graph shows the relationship between the 15 top
948 enriched BP GO terms (light pink node) and their associated gene (light blue nodes). Node size reflects
949 the number of genes associated with each GO term (Count), and edges indicate gene-GO term
950 associations. Genes in each graph include predicted gene targets of A) BTA-miRNA-DE, B) BLV-miRNA,
951 and C) BFV-miRNA, respectively.

Table 1. Sample characterization and small RNA sequencing output

| Sample_ID | PR (SD)[1] | PVL | RIN | Raw Reads (n) | Reads after QC, merging and dedupli-cation (n) | Reads mapped to bovine (%) / viral mature miRNAs[2,3] (%) | Total number of identified miRNAs (viral-miRNAs) |
|---|---|---|---|---|---|---|---|
| 5671 | 178.6(28.4) | 63,148.0 | 9.6 | 8,090,674 | 4,243,257 | 2,353,523 (96.53) / 81,744 (3.47) | 470 (10) |
| 5841 | 191.3(32.8) | 44,234.3 | 9.6 | 11,869,092 | 4,038,887 | 2,556,676 (97.08) / 74,665 (2.92) | 403 (10) |
| 6021 | 169.8(42.8) | 52,902.4 | 7.8 | 15,487,902 | 6,148,370 | 4,484,871 (92,62) / 331,369 (7.38) | 503 (10) |
| 6097 | 183.2(25.4) | 91,255.4 | 9.3 | 12,511,860 | 4,744,543 | 3,485,727 (94.31) / 198,477 (5.69) | 460 (10) |
| 5830 | <25 | - | 9.0 | 11,903,095 | 5,115,654 | 3,835,811 (99.995) / 215 (0.005) | 503 (0) |
| 6493 | <25 | - | 9.3 | 10,871,552 | 4,312,484 | 3,180,484 (99.994) / 190 (0.006) | 482 (0) |
| 6962 | <25 | - | 9.0 | 11,545,292 | 4,950,884 | 3,575,583 (99.994) / 224 (0.006) | 510 (0) |

[1]Anti-BLV enzyme-linked immunosorbent assay % of reactivity; % of reactivity values < 25.0 indicated negative samples

Abbreviatures: ID = sample identification; PR = % of reactivity; SD = standard deviation; RIN = RNA integrity number; PVL = T-3 and T0 BLV average proviral load; QC = quality control.

[2] https://www.mirbase.org/.

[3] A total of 95.7% of reads from BLV(+) samples (n = 18,402,926) and 99.9% of reads from BLV(-) samples (n = 14,378,553) aligned at least once to the Bos taurus reference genome (acc. ARS-UCD1.2). Number of reads aligned to the BLV (acc. NC_001414.1) and BFV (acc. NC_001831.1) reference genomes: 648,383 reads from all BLV(+) samples showed at least one alignment to BLV, while only 600 reads from BLV(-) cows mapped to the same genome; regarding the BFV, 138,138 reads from BLV(+) and 141 reads from BLV(-) samples had at least one aligment..

Table 2. Viral derived-miRNAs expression

| Condition | BLV(+) | | | | | BLV(-) | | | |
|---|---|---|---|---|---|---|---|---|---|
| miRNA/ID | 5671 | 5841 | 6021 | 6097 | avgexp(SD) | 5830 | 6493 | 6962 | avgexp(SD) |
| bfv-miR-BF1-5p | 8,973.2 | 719.9 | 14,813.3 | 2,872.6 | 6,844.7(6,359.3) | 15.6 | 5.8 | 15.4 | 12.3(5.6) |
| bfv-miR-BF1-3p | 12,947.2 | 1,329.6 | 15,826.4 | 2,925.9 | 8,257.3(7,204.2) | 7.1 | 11.6 | 8.6 | 9.1(2.3) |
| bfv-miR-BF2-5p | 17,694.3 | 1,735.7 | 33,383.7 | 6,868.5 | 14,920.6(13,991.2) | 0.0 | 15.4 | 17.1 | 10.8(9.4) |
| bfv-miR-BF2-3p | 2.8 | 0.0 | 6.3 | 6.2 | 3.8(3.0) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |
| blv-miR-B1-5p | 0.0 | 0.0 | 0.0 | 0.0 | 0(0.0) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |
| blv-miR-B1-3p | 7,556.4 | 13,726.1 | 46,243.5 | 43,198.3 | 27,681.1(19,875.4) | 28.3 | 24.1 | 24.0 | 25.5(2.5) |
| blv-miR-B2-5p | 30,941.9 | 35,856.1 | 88,113.4 | 81,320.4 | 59,057.9(29,825.4) | 60.9 | 69.4 | 71.9 | 67.4(5.7) |
| blv-miR-B2-3p | 34.2 | 47.4 | 171.3 | 197.0 | 112.5(83.6) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |
| blv-miR-B3-5p | 262.0 | 371.8 | 259.8 | 344.7 | 309.6(57.3) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |
| blv-miR-B3-3p | 37,074.5 | 33,573.4 | 61,540.1 | 65,060.6 | 49,312.2(16,278.9) | 39.7 | 55.9 | 53.9 | 49.8(8.8) |
| blv-miR-B4-5p | 1.4 | 0.0 | 0.0 | 1.0 | 0.6(0.7) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |
| blv-miR-B4-3p | 39.9 | 69.9 | 79.8 | 70.8 | 65.1(17.4) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |
| blv-miR-B5-5p | 864.3 | 972.1 | 1,218.4 | 752.0 | 951.7(199.2) | 0.7 | 1.0 | 0.9 | 0.8(0.1) |
| blv-miR-B5-3p | 0.0 | 1.2 | 0.8 | 2.1 | 1.0(0.9) | 0.0 | 0.0 | 0.0 | 0.0(0.0) |

Abbreviatures: ID = Sample ID; avgexp = normalized mean expression; SD = Standard deviation