Mapping regulatory networks underlying _Leishmania_ stage differentiation reveals an essential role for protein degradation in parasite development

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Keywords: _Leishmania donovani_, stage differentiation, systems analysis, protein degradation, regulatory networks

ABSTRACT

Genetic feedback control represents a central paradigm in regulation of biological systems and their

response to environmental change. Vector-borne pathogens have evolved complex developmental

programs to adapt to very distinct host environments, but the relevance of feedback regulation in stage

differentiation remains to be elucidated. Here we address this open question in the trypanosomatid

parasite *Leishmania* that shows constitutive gene transcription, thus providing a unique model system

to assess post-transcriptional mechanisms of feedback regulation in the absence of confounding

transcriptional control. Using a five-layer integrative systems analysis (from genome to metabolome),

we examined hamster-isolated *Leishmania donovani* amastigotes and culture-derived insect-stage

promastigotes. This approach enabled us to rule out genomic adaptation as a driver of parasite stage

differentiation, confirm the pivotal role of differential mRNA turnover in stage-specific gene

expression, and uncover an unexpectedly broad dynamic range of protein abundance changes that

correlated poorly with transcript levels. This discrepancy was attributed to (i) stage-specific

translational control, as indicated by alterations in snoRNA expression and changes in rRNA

modification they guide, and (ii) differential protein degradation, as revealed by quantitative proteomics

of parasites treated with the proteasomal inhibitor lactacystin. Notably, lactacystin treatment stalled the

transition of spleen-derived amastigotes into promastigotes in culture, further underscoring the role of

proteasomal activity in stage differentiation. Integration of our five-layer systems analysis established

the first link between *Leishmania* development and the expression of highly connected, stage-specific

regulatory networks encompassing mRNA turnover, protein translation, phosphorylation, and

degradation. These networks engage in complex recursive (self-regulating) interactions involved in

post-transcriptional regulation, ribosomal biogenesis, and signal transduction, as well

as reciprocal (cross-regulating) interactions, including a network between protein kinases that

phosphorylate proteasomal components in a stage-specific manner and proteasomal activities that, in

turn, target protein kinases for stage-specific degradation. Our findings provide a powerful experimental

framework to dissect the emergent properties of these regulatory feedback loops, offering critical

insights into intracellular infection and serving as a blueprint for other vector-borne pathogens that rely

on disease-associated developmental transitions.

INTRODUCTION

During their infectious cycle, many protist and fungal pathogens undergo complex developmental

transitions that adapt their biology to different host environments, thus ensuring transmission,

persistence or immune evasion. For example, stage differentiation of the malarial parasite *Plasmodium*

spp is tightly controlled at transcriptional levels by the dynamic interplay between epigenetic regulators

that remodel the chromatin landscape to accommodate stage-specific transcription factors, such as AP2

DNA-binding family members (1, 2). Likewise, the transition from fast-growing tachyzoites to slow-

growing bradyzoites during mammalian *Toxoplasma* infection is regulated by the single master TF

BFD1 (3). Unlike these apicomplexan parasites, stage-differentiation in the trypanosomatid pathogen

Leishmania spp is not regulated at the level of transcriptional control, raising the question of how these

parasites establish and maintain the various adaptive forms that develop inside their insect and

mammalian hosts.

Leishmania spp are protist parasites of humans that represent a global public health problem

causing a series of immunopathologies termed the leishmaniases (4-7). These parasites differentiate

into various developmental forms that are adapted for extracellular proliferation inside the sand fly

midgut (procyclic promastigotes), transmission from the vector to the mammalian host during uptake

of a blood meal (metacyclic promastigotes), and intracellular proliferation inside fully acidified

macrophage phagolysosomes (amastigotes) (8). Unlike other eukaryotes, *Leishmania* development is

not controlled transcriptionally as protein-coding genes in these parasites lack individual promoters and

are constitutively transcribed at all stages involving parasite-specific processes such as polycistronic

transcription and trans-splicing (9, 10). Despite the largely constitutive gene transcription, different

Leishmania stages are characterized by striking morphological, biochemical and metabolic changes

(11-16), and these are regulated by a plethora of post-transcriptional mechanisms, including differential

mRNA turn over through the binding of various proteins and protein complexes to the 5' and 3'

untranslated regions (UTRs)(17-22), that can trigger mRNA degradation via deadenylation and

decapping (23, 24). Differential gene expression in *Leishmania* can further be regulated at the

translational level, for example through (i) binding to the 5' UTR of one of the six isoforms of the

Leishmania cap-binding protein eIF4E that may confer selective ribosome recruitment, (ii) the

recruitment of Poly(A) Binding Protein 1 (PABP1) to the 3' UTR promoting translation by stabilizing

the poly(A) tail, or (iii) the differential expression of ribosomal proteins that can increase the landscape

of specialized ribosomes (25-34). Finally, differential gene expression in *Leishmania* can also be

controlled at the level of protein stability, through stage-specific expression of proteasomal components,

autophagy-related genes (ATGs) or lysosomal proteases (35-41).

While each level of gene expression control has been studied individually in different

Leishmania species, how these different types of regulation are integrated during *Leishmania* stage

differentiation remains to be elucidated. Here, we address this open question by applying a 5-layer,

systems level analysis on the two major life cycle stages of *Leishmania donovani*, i.e. hamster-isolated,

bona-fide amastigotes and derived promastigotes. Our analysis draws a complex picture of the

Leishmania differentiation process that emerges from the interactions between recursive (i.e. self-

controlling) and reciprocal (i.e. cross-controlling) feedback loops between mRNA turnover, protein

translation, protein phosphorylation and protein degradation. Our data reveal *Leishmania* as an

interesting model system to analyze phenotypic adaptation in the absence of transcriptional control and

to dissociate the role of complex regulatory networks in the development of stable, disease-causing life

cycle stages in eukaryotic pathogens.

MATERIAL & METHODS

Ethics statement. Work on animals was performed in compliance with French and European regulations on care and protection of laboratory animals (EC Directive 2010/63, French Law 2013-118, February 6th, 2013). All animal experiments were approved by the Ethics Committee and the Animal welfare body of Institut Pasteur and by the Ministère de l'Enseignement Supérieur, de la Recherche et de l'Innovation (project n°#19683).

Animals. Twenty one female Golden Syrian hamsters (*Mesocricetus auratus* RjHan:AURA, weighting between 50 – 60 g) were purchased from Janvier Laboratories. All animals were handled under specific, pathogen-free conditions in biohazard level 3 animal facilities (A3) accredited by the French Ministry of Agriculture for performing experiments on live rodents (agreement A75-15-01).

Parasites and culture. *Leishmania donovani* strain 1S2D (MHOM/SD/62/1S-CL2D) was obtained from Henry Murray, Weill Cornell Medical College, New York, USA and maintained by serial passages in hamsters. Anesthetized hamsters were inoculated by intra-cardiac injection of $5x10^7$ amastigotes purified from hamster infected spleens as described previously (42). The weight of the animals was recorded over time, and the animals were euthanized by CO_2 asphyxiation before they reached the end-point of infection represented by a 20% loss of body weight. Amastigotes were then recovered from the infected hamster spleens and used for nucleic acid and protein extractions, or differentiated into promastigotes at 26°C in M199 complete medium (M199, 10% FBS, 25 mM HEPES; 100 μM adenine, 2 mM L-glutamine, 10 μg/ml folic acid, 13.7 μM hemin, 4.2 mM $NaHCO_3$, 1xRPMI1640 vitamins, 8 μM 6-biopterin, 100 units penicillin and 100 μg/ml streptomycin, pH 7.4). Promastigotes, derived from splenic amastigotes, were collected after 2 *in vitro* passages for nucleic acid and protein extractions.

Experimental design. Strains issued from independent experimental evolution assays are identified by number, e.g. amastigote ama1 and promastigote pro1 are the parasites prepared from hamster H1 (see Figures 1A and S1 for details). For comparative analyses, DNA, RNA, proteins and metabolites were

extracted from different splenic ama and their matching pro parasites, the latter being collected for

extraction during exponential growth phase.

Parasite differentiation and lactacystin treatment. Spleen-derived amastigotes were incubated at

$2x10^7$/ml and 26°C in presence or absence of 10 μM of lactacystin (L6785, Sigma) for 3, 6, 18 and 48h

in M199 complete medium. At each time point, parasites were collected to control for the presence of

paraflagellar rod protein 2 (PFR2) as promastigote differentiation marker. After one *in vitro* passage,

amastigote-derived promastigotes were incubated at a concentration of $2x10^7$ parasites per ml in

presence or absence of 10 μM of lactacystin for 18h. Parasite viability was assessed by FACS analysis

after 18h of treatment by propidium iodide (1μg/ml) or YO-PRO-1 (0.2 μM) (# Y3603, Invitrogen)

staining as readout. Parasites were collected from 25 ml of culture after 18h of treatment for

quantitative, label-free proteomics analyses.

Microscopy. Differential interference contrast (DIC) microscopy was applied on hamster-derived

amastigotes during *in vitro* differentiation to promastigotes using an Axioplan 2 imaging microscope

using a 63x oil immersion objective, the AxioVision Rel.4.8 software and an AxioCam MRm camera

(Carl Zeiss). Alternatively, microscopic images of amastigotes in presence or absence of the lactacystin

inhibitor were obtained using the EVOS FL microscope (life technologies) at a 20x magnification.

Western blot analysis. Protein extracts were obtained from parasites after 3, 6, 18 and 48h of

differentiation from amastigotes to promastigotes in presence or absence of lactacystin. Extracts were

loaded on 4-12% SDS-PAGE (NuPage 4-12% Bis-Tris gel NP0321BOX, Invitrogen), separated by

electrophoresis and stained with Coomassie blue or transferred onto PVDF membrane. The presence of

PFR2 and α-tubulin was revealed using the anti-PFR2 antibody (kindly given by Philippe Bastin), an

anti- α-tubulin antibody (clone B-5-1-2, Sigma), the corresponding HRP-conjugated secondary

antibodies (Invitrogen) and the SuperSignal West Pico PLUS kit (Thermoscientific).

Genome sequencing and data analysis. DNA was prepared from splenic amastigotes and

promastigotes at exponential culture phase. Parasites were centrifuged at 1,600 x *g* (pro) or 2,000 x *g*

(ama) for 10 min at room temperature. Approximately $1x10^8$ promastigotes from logarithmic growth

phase were re-suspended in 200 µl PBS and genomic DNA was purified using DNeasy Blood and

Tissue kit from Qiagen and RNase A according to the manufacturer's instructions. DNA concentrations

were measured in duplicate by fluorescence using a Molecular Device fluorescence plate reader (Quant-

IT kits, Thermo Fischer Scientific). The quality of DNA was controlled by determining the DNA

Integrity Number (DIN) analyzing 20 ng of DNA on a TapeStation 4200 (Agilent). One µg of genomic

DNA was used to prepare a library for whole genome sequencing on an automated platform, using the

Illumina "TruSeq DNA PCR-Free Library Preparation Kit", according to the manufacturer's

instructions. After normalization and quality control, qualified libraries were sequenced on a HiSeqX5

platform from Illumina (Illumina Inc., CA, USA) at the Centre National de Recherche en Génétique

Humaine (CEA, Evry, France), generating paired-ended, 150-bp reads. Sequence quality parameters

were assessed throughout the sequencing run. Standard bioinformatics analysis of sequencing data was

based on the Illumina pipeline to generate a FASTQ file for each sample.

Genomic DNA reads were aligned to the *L. donovani* Ld1S reference genome

(https://www.ncbi.nlm.nih.gov/bioproject/PRJNA396645, GCA_002243465.1) with BWA mem

(version 0.7.12) with the flag -M to mark shorter split hits as secondary. Samtools fixmate, sort, and

index (version 1.3) were used to process the alignment files and turn them into bam format (43).

RealignerTargetCreator and IndelRealigner from the GATK suite were run to homogenize indels (44).

Eventually, PCR and optical duplicates were labeled with Picard MarkDuplicates (version 1.94(1484))

(https://broadinstitute.github.io/picard/) using the option

"VALIDATION_STRINGENCY=LENIENT". For each read alignment file, Samtools view (version

1.3) and BEDTools genomecov (version 2.25.0) were used to measure the sequencing depth of each

nucleotide (45). Samtools was run with options "-q 50 -F 1028" to discard reads with a low map quality

score or potential duplicates, while BEDTools genomecov was run with options "-d -split" to compute

the coverage of each nucleotide. The coverage of each nucleotide was divided by the median genomic

coverage. This normalization is done to account for library size differences. The chromosome

sequencing coverage was used to evaluate aneuploidy between EP.1 and LP.1 samples. Then for each

sample and for each chromosome, the median sequencing coverage was computed for contiguous

windows of 2,500 bases. As previously published (46), the stably disomic chromosome 36 was used to

normalize chromosome read depth and to estimate chromosome polysomy levels in each sample. Gene

counts were produced using featureCounts (version 1.4.6-p3 (47)) with these parameters: -s 0 -t gene -

g gene_id and were normalized according to the median-ratio method.

For the generation of the chromosome median somy score heatmap, mean coverage in 300 bp bins as

generated by the GIP pipeline were used to compute somy scores per chromosome by first normalizing

bin scores for a sample by their median across all genome (to obtain comparable values between

samples), multiplying by two (to scale somy values to the default diploid state assumed for most of the

genome), and then taking the median across bins belonging to a given chromosome. These chromosome

scores were then displayed on a sample x chromosome heatmap, excluding the "maxi" chromosome.

This was done using the Pandas 1.4.2 {McKinney, 2010} (Reback *et. al*, 2021. Pandas-dev/pandas:

Pandas 1.2.4, https://zenodo.org/records/4681666)**,** Matplotlib 3.5.1 {Hunter, 2007} and Seaborn

0.11.2 {Waskom, 2021} Python libraries.

For the assessment of gene coverage ratio, normalized mean coverages per gene are reported

by GIP (version 1.1.0 (48)), as the mean coverage of the gene divided by the median coverage of the

chromosome containing the gene. For a given gene, and a given amastigote - promastigote pair, a gene

coverage ratio was computed adding 0.1 to the gene coverages of the individual samples, in order to

avoid zero division errors. The genes were sorted by genomic coordinate and assigned successive

integers to create a "genomic index". Gene coverage ratios were then plotted along this genomic index,

using the same Python libraries as above.

For the SNP analysis, alternative (Alt) allele frequencies for SNPs were retrieved from the

output of the SNV giptools module (provided along with the GIP pipeline). Euclidean distances between

samples in the SNP allele frequencies space were computed and used to build a dendrogram using the

scipy.cluster.hierarchy Python module (SciPy version: 1.8.0). The clustering was obtained using the

UPGMA method.

RNA-seq analysis. Total RNA was extracted from splenic amastigotes and promastigotes in

exponential culture phase. Amastigotes and promastigotes were centrifuged respectively at 2,000 x g or

1,600 x g for 10 min at 4°C, and re-suspended in the lysis buffer supplied with the Qiagen RNeasy Plus

kit. The samples were stored at -80°C and RNA extractions were performed according to the

manufacturer's instructions, including a DNase treatment. RNA integrity was controlled using the

Agilent Bioanalyzer. DNase-treated RNA extracts were used for library preparation using the TruSeq

Stranded mRNA sample preparation kit (Illumina, San Diego, California) according to the

manufacturer's instructions. An initial poly (A+) RNA isolation step (included in the Illumina protocol)

was performed on total RNA to remove ribosomal RNA. Fragmentation was performed on the enriched

fraction by divalent ions at high temperature. The fragmented RNA samples were randomly primed for

reverse transcription followed by second-strand synthesis to create double-stranded cDNA fragments.

No end repair step was necessary. An adenine was added to the 3'-end and specific Illumina adapters

were ligated. Ligation products were submitted to PCR amplification. The obtained oriented libraries

were controlled by Bioanalyzer DNA1000 Chips (Agilent, # 5067-1504) and quantified by

spectrofluorimetry (Quant-iT High-Sensitivity DNA Assay Kit, #Q33120, Invitrogen). Sequencing was

performed on the Illumina Hiseq2500 platform at the Biomics Center (Institut Pasteur, Paris, France)

to generate single-ended, 130-bp reads bearing strand specificity.

The bioinformatics analysis was performed using the RNA-seq pipeline 'Sequana' (49). Reads

were cleaned of adapter sequences and low-quality sequences using cutadapt (version 1.11) (50). Only

sequences of at least 25 nucleotides in length were considered for further analysis. STAR (version

2.5.0a) (51), with default parameters, was used for alignment on the reference genome

(GCA_002243465.1). Genes were counted using featureCounts (version 1.4.6-p3) (47) from Subreads

package (parameters: -t gene -g gene_id -s 1). Count data were analyzed using R version 3.6.1 (52) and

the Bioconductor package DESeq2 (version 1.24.0) (53). The normalization and dispersion estimation

were performed with DESeq2 using the default parameters and statistical tests for differential

expression were performed applying the independent filtering algorithm. For each pairwise comparison,

raw p-values were adjusted for multiple testing according to the Benjamini and Hochberg (BH)

procedure (54) and genes with an adjusted p-value lower than 0.01 were considered differentially

expressed.

Label-free, quantitative total proteome analyses. Untreated amastigotes (ama and ama-18h) and

promastigotes (pro), and lactacystin-treated parasites (ama lacta and pro lacta) were recovered and

washed three times with cold PBS at 2,000 x g or 1,600 x g for 10 min at 4°C. Parasite lysates were

prepared in eFASP lysis buffer (4% SDS / 0,2 % DCA / 50mM TCEP / 50 mM ammonium bicarbonate

buffer pH 8) (55). For detailed protocol see supplementary data file. MS scans were acquired at a

resolution of 70,000 and MS/MS scans (fixed first mass 100 m/z) at a resolution of 17,500. The AGC

target and maximum injection time for the survey scans and the MS/MS scans were set to 3 x 10^6 for

20 ms and 10^6 for 60 ms, respectively. An automatic selection of the 10 most intense precursor ions was

activated (Top 10) with a 40 s dynamic exclusion. The isolation window was set to 1.6 m/z and

normalized collision energy fixed to 28 for HCD fragmentation. We used a minimum AGC target of

10^4 corresponding to an intensity threshold of 1.7 x 10^5. Unassigned precursor ion charge states as well

as 1, 7, 8 and >8 charged states were rejected and peptide match was disable.

Label-free quantitative phosphoproteome analyses. Amastigote (ama) and promastigote (pro)

parasites were recovered and washed three times by centrifugation in cold M199 at respectively 2,000

x g or 1,600 x g for 10 min at 4°C. Samples were incubated for 10 min at 4°C in lysis buffer (1ml per

1.5×10^9 parasites) consisting of 8 M urea, 50 mM Tris, supplemented with a protease inhibitor cocktail

(cOmplete from Roche) and a phosphatase inhibitor cocktail (PhosStop from Roche).   Following

sonication for 5 min using a sequence of 10 s pulse and 20 s pause, the lysates were centrifuged 15 min

at 14,000 x g and 4°C, and the supernatant was collected and stored at -80°C until use. Proteins were

quantified by RC DC protein assay (Bio-Rad) and adjusted to 1.3 µg.µl^{-1} in lysis buffer. For detailed

protocol see supplementary data file. Phosphopeptide enrichment was carried out as described in (56)

and detailed in supplementary data. All analyses were performed on a Q Exactive HF Mass

Spectrometer (Thermo Fisher Scientific) coupled with a Proxeon EASY-nLC 1000 (Thermo Fisher

Scientific) as detailed in supplementary data. Details for the proteomics and phosphoproteomics

analyses following the data acquisition are available in supplementary data. Results are presented in

table 6 and 10.

Phosphopeptides were selected for Gene Ontology enrichment analyses if (i) they were

exclusively identified in one of the two stages, (ii) they showed significant, stage-specific changes in

phosphopeptide abundance (fold change ≥ 2, adj. p-value < 0.01) even if the corresponding protein was

not detected in the total proteome analysis, and (iii) they showed a significant increase in

phosphostochiometry (p-value < 0.05) as calculated by the ratio 'change in phosphopeptide abundance'

vs 'change in protein abundance' using a cut off of fold changes ≥ 2 and adj. p-value < 0.01 for both

analyses (see table 11).

Metabolomic analysis. The sample extraction was performed as described previously (57). Briefly, 10^8

cells were used per each 200 μL sample. Firstly, cells were rapidly quenched in a dry ice/ethanol bath

to 4°C, then centrifuged, washed with 1 x PBS, and resuspended in the extraction solvent

(chloroform:methanol:water, 1:3:1). After vigorous shaking at 4°C for 1 h, extracts were centrifuged

(16,000 x g, 4°C, 10 min) and the supernatants collected and stored at -80°C until the analysis. LC-MS

analyses were performed using separation on 150 x 4.6 mm, 5 mm ZIC-pHILIC (Merck) on UltiMate

3000 RSLC (Thermo Scientific) followed by mass detection on an Orbitrap Exactive mass spectrometer

(Thermo Fisher) at Glasgow Polyomics. Analyses were performed in positive and negative polarity

switching mode, using 10 μl injection volume and a flow rate of 300 μl/min over 26 min on the column

maintained at 30°C, as follows: 0 to 20 min 20%-to 80% solution A, 15 to 17 min 95% solution A, 17

to 26 min 20% solution A where solution A is 20 mM ammonium carbonate in water and solution B is

acetonitrile). The samples were run alongside 249 authentic standards at 10 μM each. Mass

spectrometry data was processed using Mzmatch (58) and Ideom (57) software. Unique signals were

extracted using the centwave algorithm and matched across biological replicates based on mass-to-

charge ratio and retention time. These grouped peaks were then filtered based on relative standard

deviation and combined into a single file. The combined sets were then filtered on signal-to-noise score,

minimum intensity and minimum detections. The final peak set was then gap-filled and converted to

text for use Putative metabolite identification corresponds for the most part to Metabolite Standards

Initiative (MSI) level 2 (mass only, and thus only considered a tentative annotation), whereas

metabolites matching in retention time to an included standard correspond to level 1 (considered likely

an accurate annotation). Peaks having an area with root squared deviation across pooled samples > 50%

were excluded, as were those with a retention time < 4 min (due to poor resolution).

Quantification of pre-snoRNA levels. DeSeq2 analysis was performed on two biological replicates of

amastigote and promastigote poly(A)-RNA libraries prepared as previously described (34). Fold change

> 1.5 and adj. p-value < 0.05 was considered as significantly upregulated RNA and Fold change <1.5,

adj. p-value < 0.05 was considered as significantly downregulated RNA.

Localizing rRNA modification on Leishmania cryo-EM ribosome structure. The cryo-EM atomic

model of *Leishmania major* 80S ribosomes bound to mRNA and all three tRNAs (PDB: 8RXH) was

used to indicate rRNA modifications of the ribosome. Figures were generated using UCSF Chimera-X

software (59).

Systems-level analyses. A lists of *L. donovani* GO terms (60) was built in house. The Biological

Networks Gene Ontology tool (BiNGO) plugin of the Cytoscape software package (version 3.8.2) was

used to map and visualize functional enrichments in each data set based on the GO hierarchy. A

Benjamini & Hochberg false discovery rate with a significance level of 0.05 was applied. Cluster

efficiency represents the percentage of genes for a given GO term compared to the total number of

genes with any GO annotation in the considered set of genes. Enrichment score corresponds to the

percentage of genes for a given GO term compared to all the genes sharing the same GO term in the

genome. Word Cloud of GO enrichment analysis limited to GO Slim terms and a threshold of p-value

< 0.05 for the category Biological Process was performed on TriTrypdb (https://tritrypdb.org) using the

L. donovani LdBPK orthologs for all genes that showed statistically significant expression changes (p-

value < 0.01) at both RNA or protein levels.

For the phosphoproteomic work, phosphopeptides were selected for Gene Ontology enrichment

analyses if (i) they were exclusively identified in one of the two stages, (ii) they showed significant,

stage-specific changes in phosphopeptide abundance (fold change $\geq$ 2, adj. p-value < 0.01) even if the

corresponding protein was not detected in the total proteome analysis, and (iii) they showed a significant

increase in phosphostochiometry (p-value < 0.05) as calculated by the ratio 'change in phosphopeptide

abundance' vs 'change in protein abundance' using a cut off of fold changes $\geq$ 2 and adj. p-value < 0.01

for both analyses (see table 11). Functional enrichment networks were built with the STRING plugin

of the Cytoscape software package (version 3.8.2) using the *L. infantum* orthologs and a Benjamini &

Hochberg p-value threshold < 0.05. The "ClusterProfiler" package (version 4.2.2) of R was used for

KEGG (Kyoto encyclopedia of genes and genomes) gene set enrichment analysis (GSEA) and for data

mapping on metabolic pathways available in the KEGG database. Results were visualized using the

"pathview" packages (version 1.34.0).

Data availability. The mass spectrometry proteomics data were deposited to the ProteomeXchange

Consortium via the PRIDE partner repository with the dataset identifiers PXD035697

(phosphoproteome data) and PXD035698 (proteome upon lactacystin treatment) (61). The DNAseq

data have been deposited in NCBI's Gene Expression Omnibus (62) PRJNA1231373 and are

accessible in the BioProject PRJNA1231373 (https://www.ncbi.nlm.nih.gov/sra/PRJNA1231373).

The RNAseq data have been deposited in NCBI's Gene Expression Omnibus (62) and are accessible

through GEO Series accession number GSE227637

(https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc= GSE227637). The mass spectrometry

metabolomic data were deposited in the Metabolights data repository.

RESULTS

***Leishmania* stage differentiation occurs independently from changes in gene dosage.** We carried

out an in-depth, systems analysis of independent *L. donovani* amastigote preparations isolated from

infected hamster spleens and their culture-derived promastigotes at *in vitro* passage 2, with the aim of

revealing mechanisms controlling stage-specific differential expression across five quantifiable information levels, by comparing the genome, transcriptome, proteome, metabolome and phospho-proteome of the two life-cycle stages (Figure 1A and Figure S1). We first investigated the possible role of genome instability in stage-differentiation, given the very well documented propensity of *L. donovani* to establish and tolerate chromosome and gene copy number variations in culture and during sand fly or hamster infection (46, 60, 63).

Applying our computational pipeline termed 'GIP' (48) to the genome sequences of three independent pairs of amastigotes and promastigotes confirmed the largely disomic state of the *Leishmania* genome at both stages, except for the known constitutive tetrasomy of chr 31 (Figure 1B and Figure S2A, Table 1). While we did not detect any major changes at karyotypic levels, a clear pattern emerged for promastigotes that showed a slight but reproducible increase in median coverage for chr 1, 2, 3, 4, 6 and 20, suggesting the emergence of mosaic aneuploidy for these chromosomes early during culture adaptation. Likewise, stage differentiation was not associated with any major gene copy number variation (Figure 1C, Table 2). Even though a number of genes showed increased (e.g. Ld1S_300477100 encoding an intraflagellar transport (IFT) protein) or decreased (e.g. Ld1S_220235700 encoding a hypothetical protein) read depth in individual pro strains, none of these changes converged across all pro samples, suggesting that they may either represent intrinsic variations between different amastigote isolates or are the result of random genetic drift, likely caused by the bottle neck effect represented by culture inoculation. Plotting the normalized read depth ratios between amastigotes and promastigotes revealed an intriguing, undulating pattern that was highly reproducible across the three independent experiments and has been previously associated with nascent DNA produced during the replicative S-phase of the cell cycle (64-66). Finally, differentiation did not affect the SNP frequency distribution (Figure S2B, Table 3), which resulted in a largely diagonal pattern when plotting amastigote against promastigote SNP frequencies (Figure 1D).

In conclusion, in contrast to fitness gain (46, 67), *L. donovani* promastigote differentiation *in vitro* is independent of genomic adaptation, confirming our previous results obtained with sand fly-derived parasites (60, 68).

Comparative RNA-seq analysis reveals stage-specific regulons that are controlled at post-transcriptional levels. We next analyzed stage-specific expression changes applying RNA-seq analyses on two amastigote and derived promastigote strains. In contrast to the relatively stable genome infrastructure maintained between stages, we observed substantial, stage-specific differences in transcript abundance. From 8850 detected transcripts in both stages, 4998 showed statistically significant differences (adj. p-value < 0.01), including 531 and 1082 transcripts with a 2-fold or higher increase in abundance in amastigotes and promastigotes, respectively (Figures 2B left panel,) (Table 4). Considering the lack of promoter-driven control of gene expression in *Leishmania* and the absence of stage-specific gene dosage effects (see Figure 1), the observed differences in amastigote and promastigote transcript abundances are most likely attributable to stage-specific mRNA turnover. Direct comparison of replicate transcriptomes did not identify any significant differences between the two independent amastigote and promastigote expression profiles, revealing the high fidelity of post-transcriptional mechanisms that govern stage-specific mRNA turnover (Figure 2A).

Analyzing the stage-specific RNA-seq data sets identified clusters of co-regulated transcripts that share a common annotation, which therefore define distinct regulons (Figure 2B, Table 4). In amastigotes, one major positive regulon showing increased abundance included 83 amastin transcripts encoded on 5 different chromosomes, confirming the post-transcriptional co-regulation of this gene family as previously described (19, 69, 70) (Table 4, sheet C and E). Another positive regulon in amastigotes was associated with protein translation, with GO enrichment observed for the terms 'ribosome biogenesis' (38 transcripts), 'rRNA processing' (27 transcripts), 'tRNA processing' (7 transcripts), or 'RNA methylation' (5 transcripts) (Figure 2B, middle panel). These data reveal the ribosome and epi-transcriptomic regulation as possible central components in *Leishmania* stage differentiation.

In promastigotes, a first positive regulon included 23 transcripts annotated for the GO term 'ciliary cell motility', confirming the post-transcriptional co-regulation of flagellar biosynthesis as previously shown in *L. mexicana* (71) (Figure 2B right panel, Table 4, sheet F). Further regulons are defined by the GO terms 'oxidative phosphorylation' (12 transcripts), 'oxidation reduction' (46 transcripts), and 'translational elongation' (28 transcripts), the latter suggesting translational control as

a further regulatory step in stage differentiation. Finally, the largest promastigote-specific regulon is

represented by 53 transcripts annotated for 'post-transcriptional regulation of gene expression'.

In conclusion, our RNAseq data reveal a series of stage-specific regulons whose mRNA

stability is regulated in a coordinated manner. Whether this co-regulation is governed by stage-specific

mRNA decay mechanisms acting on 5' and 3' UTRs or by stage-specific mRNA stabilization via RNA-

binding proteins or mRNA modification remains to be established.

Quantitative proteome analysis reveals regulons that are independent of stage-specific

mRNA turnover. We next employed quantitative proteomic analysis to assess how the different

regulons observed in our RNAseq analyses translate into stage-specific proteomes. Label-free

quantitative proteomic analysis of 4 independent biological replicates of amastigotes and derived

promastigotes identified over 4000 proteins at each stage in at least 3 out of 4 biological replicates,

representing 3521 differentially expressed proteins (adjusted p-value < 0.01), 1534 of which were

exclusively detected in either ama or pro (Figure 3A left panel, Table 6). Considering the 822 proteins

that are specific or more abundant in amastigotes (fold change $\geq$ 2), functional enrichment was mainly

observed for the GO terms 'oxidation reduction' (61 proteins), 'vesicle mediated transport' (23 proteins)

and 'phosphorylation' (41 proteins) (Figure 3A, middle panel, Table 6, sheets B and D). Considering

the 1637 proteins that are specific or more abundant in promastigotes (fold change $\geq$ 2), functional

enrichment was observed for the GO terms 'post-transcriptional regulation of gene expression' (71),

'translation' (69 proteins), 'ribosome biogenesis' (21 proteins) and 'ciliary cell motility' (42 proteins)

(Figure 3A, right panel, Table 6, sheet C and E). The metabolomic analysis of four independent

amastigote and promastigote samples (Figure S4A) support our proteomic data. Promastigotes showed

increased amino acid biosynthetic activity that matched the changes in protein abundance observed for

this pathway, while the amastigote metabolome indicates a shift from glycolytic to TCA cycle-

dependent energy production that was previously described (12, 14, 72) (Figures S4B-D), with a notable

increase in metabolites involved in butanoate metabolism indicative of a shift between carbohydrate

use as a primary energy source to utilization of fatty acid oxidation. While succinate and malate levels

were diminished this is because these metabolites comprise a glycosomal succinate shunt related to

glycolysis in addition to their roles in the TCA cycle. The amastigote expression profile and metabolome further indicate the stage-specific production of the storage-carbohydrate mannogen, thus confirming previous observations (73) (Figure S4E). Surprisingly, only few enriched GO terms correlated between transcriptomics and proteomics datasets, suggesting additional layers of regulation between the transcriptional and translational levels of gene expression. To investigate this aspect on a genome-wide level, we next plotted the simple numerical values we obtained when calculating the ratios of expression changes between ama and pro observed in the transcriptomics and proteomics datasets (Figure 3B, Table 7). The double ratio analysis of 2076 genes showing statistically significant expression changes at both mRNA and protein levels (p-value < 0.01) defined four quadrants: Quadrants 2 (Q2) and 4 (Q4) correspond to 365 and 918 genes (Table 7, sheet E and I), respectively, for which stage-specific expression changes correlate between protein and RNA levels, suggesting that the abundance of these proteins is mainly regulated by mRNA turn-over. Q2 (mRNA and protein up in ama) includes the GO terms 'vesicle mediated transport' (14 genes) and 'lipid metabolic process' (18 genes) (Table 7, sheet F), while Q4 (mRNA and protein up in pro) includes the GO term 'cell motility' (41 genes) (Table 7, sheet J).

Counter-correlations between RNA and protein abundances (p-value < 0.01) were observed for 381 and 412 genes, respectively falling into quadrants Q1 (RNA down and protein up in ama) and Q3 (RNA down and protein up in pro) (Figure 3B, Table 7, sheets C and G). The most important difference was observed for 38 genes annotated for the GO term 'ribosome biogenesis' (Figure 3C, Table 7, sheet H). While this GO term defined a positive regulon at mRNA level in amastigotes (see Figure 2B), the same genes show reduced expression at this stage at protein level (Figure 3C). This lack of correlation may be explained by stage-specific changes in mRNA translation efficiency. While it was technically impossible to enrich enough material from spleen-derived amastigotes to perform in-depth ribo-profiling analyses, we obtained indirect evidence for differential translational control: First, analyzing poly-A transcriptome libraries of amastigotes and promastigotes we observed significant differences in stage-specific expression of 110 pre-snoRNAs (Table 5, Figure 3D). These data fit with the GO enrichment 'ribosome biogenesis' observed in the promastigote proteome and the expression of a series of methyl- and pseudouridyl-transferases at this stage (see Table 6, sheets A and E). Second, by

localizing RNA modification on high resolution cryo-EM structure of *Leishmania* ribosome we indeed

observed important stage-specific changes in Pseudouridine and 2'-O-Methylation rRNA modifications

(Figure 3 E), which are known to affect the ribosome translational activity (34).

In conclusion, the discrepancies observed between transcript and protein abundances in *L.*

donovani amastigotes and promastigotes reveals stage-specific regulatory mechanisms independent of

mRNA stability, which may act at the level of translational control, or rely on differential protein

stability as investigated below.

Stage-specific protein turn-over is required for parasite differentiation. The poor correlation we

observed between changes in mRNA and protein abundance during the amastigote-to-promastigote

transition (see Figure 3B) primed us to investigate the role of proteasomal protein degradation in

Leishmania differentiation by employing the highly specific and irreversible proteasome inhibitor

lactacystin (74). Comparative, quantitative proteomics analyses were performed with (i) spleen-derived

amastigotes (ama), (ii) amastigotes after the first 18h of differentiation in the absence (control, ama-

18h) and the presence of latacystin (ama-lacta), and (iii) promastigotes in culture without (pro) and after

18h of treatment with lactacystin (pro-lacta) (Table 8, Figure 4A and S5).

First, this data set allowed us important insight into processes of early differentiation, revealing

increased abundance in ama-18h compared to ama for 694 proteins (fold change ≥ 2, adjusted p-value

< 0.01) (Table 8b, sheet B) with GO enrichment observed for the terms 'ribosome biogenesis', 'RNA

methylation', 'rRNA processing', 'gene expression', and 'post-transcriptional regulation of gene

expression', the latter including 6 genes encoding for pseudouridine transferases known to modify

ribosomal (r) RNA (Figures S5A and S6) (Table 8b, sheets D to H). These data further sustain a role of

epitranscriptomic regulation and stage-adapted ribosomes as key processes that may initiate *Leishmania*

promastigote differentiation, at least in culture.

Second, lactacystin treatment uncovered an essential regulatory role of proteasomal protein

degradation in *L. donovani* development. While the presence of the inhibitor neither affected parasite

viability (Figure S7A) nor promastigote growth (data not shown), lactacystin treatment abrogated the

amastigote-to-promastigote developmental transition as judged by the persistence of an amastigote-like

morphology (i.e. oval cell shape, retracted flagellum), and the absence of PFR2 expression

characteristic of promastigotes (Figures 4B and C). In contrast, untreated amastigotes spontaneously

converted to flagellate promastigotes after 48h in culture (Figure S7B and C).

Finally, comparing the proteomics signatures of lactacystin-treated samples and untreated

controls (Figure S5) allowed us to reveal the proteasomal targets, whose rescue correlated with the

observed abrogation of differentiation. This analysis is complicated by the fact that amastigotes will not

only respond to the 18h treatment but also initiate their differentiation into promastigotes during this

period. To distinguish changes caused by lactacysin treatment from changes due to the differentiation,

we plotted the ratio of ama/ama-lacta against the ratio ama-18h/ama-lacta (Figure 4D), which defined

one quadrant (Q3) representing 129 proteins whose increased abundance in ama-lacta was exclusively

due to rescue from degradation (Table 9, sheet A). Applying STRING and GO analyses on these

proteins revealed enrichment for the terms 'protein kinase', 'cellular response to stimulus', and 'cell

cycle' (Figure S8 lower panel, Table 9, sheet B and G).

Proteomics analysis of lactacystin-treated promastigotes (pro-lacta) revealed the rescue of 289

proteins from degradation when compared to untreated control (pro) (Table 9, sheet C). Comparison of

rescued proteins across both ama and pro stages revealed a series of intriguing findings: First, most of

these proteins were rescued in a stage-specific manner, confirming differential protein degradation as

an important new regulatory component in the development and maintenance of *Leishmania* life cycle

stages as suggested by previous reports (37, 75-81). Second, just like the proteins rescued from

proteasomal degradation in ama-lacta, the pro-lacta data set too showed GO enrichment for the terms

'protein kinase' and 'cellular response to stimulus', involving distinct sets of proteins (Figure S8, upper

panel) (Table 9, sheets D, E and G). Thus, rather than being controlled by a single master regulator,

Leishmania differentiation seems to be governed by complex signaling networks that are established in

a stage-specific manner involving proteasomal degradation. Third, when mapping the lactacystin-

rescued proteins to the established, stage-specific proteomes (Figure 4E), an intriguing pattern emerged:

the majority of proteins that were rescued in ama-lacta turned out to be expressed specifically in the

promastigote proteome. *Vice versa,* many proteins rescued in pro-lacta were part of the amastigote-

specific proteome (Table 9, sheets H and I), a pattern that affected 14 protein kinases (Figure 4F). Thus,

these proteins are expressed constitutively across both stages, while their steady-state stage-specific

abundance is regulated by proteasomal degradation. Surprisingly, 12 proteins (including the protein

kinase Ld1S_220237000) were rescued from degradation in both ama-lacta and pro-lacta but were

never identified in the proteomes of the corresponding untreated controls, suggesting that they are

constitutively degraded in both stages under our experimental conditions (Figure 4G, Table 9, sheet F).

In conclusion, our results reveal an unexpected but substantial role of protein degradation in

regulating the *L. donovani* amastigote and promastigote proteomes and identify protein kinases as key

targets of stage-specific proteasomal activities.

Quantitative phosphoproteome analysis reveals *Leishmania* reciprocal and recursive regulatory

networks underlying stage differentiation. The investigations above associated the expression of

protein kinases (and their degradation) to *Leishmania* stage differentiation (Figure 4F, S8 and S9), thus

confirming previous reports on the importance of phosphotransferase activities in trypanosomatid

development (82-92). Differential turnover of these important signaling proteins by stage-specific

proteasomal activities likely represents a regulatory switch controlling parasite development. To further

investigate how this putative regulatory switch operates, we screened for stage-specific protein kinase

substrates applying label-free, quantitative phospho-proteomic analyses on three independent

amastigote and promastigote samples. We identified a total of 2079 phospho-peptides in amastigotes

and 7095 phospho-peptides in promastigotes, suggesting a three-fold increase in global protein

phosphorylation in insect-stage parasites (Table 10). However, increased TiO_2 phosphopeptide

enrichment in a given sample may be due either to increased substrate expression or increased *de novo*

phosphorylation (Figure 5A and S10). To distinguish between these two possibilities and assess the

stage-specific phospho-stoichiometry, we normalized changes in protein phosphorylation against

changes in protein abundance (Figure 5B).

We identified 877 phosphoproteins in amastigote samples carrying 2079 phosphosites of which

55.8% (1160) were uniquely identified at this stage, while 12.6% (565) showed a significant change in

phospho-stoichiometry compared to promastigotes (Figure S10, Table 11 sheets A to D). Conversely,

2054 phosphoproteins with 7095 phosphorylation sites were identified in promastigote samples, of

which 70% (4967) were exclusively detected at this stage, while 1% (73) presented a significant change in phospho-stoichiometry compared to amastigotes (Figure S10, Table 11 sheets F to I). These data reveal a surprising level of stage-specific phosphorylation in promastigotes, which may reflect their increased biosynthetic and proliferative activities compared to amastigotes.

We next investigated the pathways regulated by phosphotransferase activities analyzing the phosphoproteomes of amastigote and promastigote for enriched biological processes (Figure 5C) (Table 11, sheets E and J). Stage-specific enrichment was observed in the pro dataset for 'ciliary cell motility', indicating that flagellar biogenesis and activity are not only regulated by increased protein abundance, but also by increased phospho-stoichiometry of flagellar proteins. Surprisingly, similar functions were enriched in both the differential datasets linked to cell division (e.g. 'cytokinesis', 'cell cycle'), signaling (e.g. 'protein aa phosphorylation', 'signal transduction'), or post-transcriptional regulation of gene expression. Thus, aside stage-specific protein turnover (see above, Figure 4), differential phosphorylation represents yet another layer of regulation that contributes to the establishment of stage-specific biological networks in *Leishmania*.

Analyzing stage-enriched, functional networks in further detail using the STRING applications of the Cytoscape software package revealed a series of intriguing regulatory interactions (Figure S11, Table 12): First, a group of protein kinase substrates are protein kinases themselves with more than 30 and 100 members of this large protein family identified in the ama and pro phosphoproteomes, respectively (Table 11, sheets D and H, see Figure 5D for phosphorylation pattern of some members of the MAP kinase pathway). Second, both amastigote and promastigote phosphoproteomes include stage-specific phosphatases (18 in ama and 55 in pro) that can counteract protein kinase activities and even deactivate signaling cascades by dephosphorylation of protein kinase substrates (Figure S11, Table 11, sheets D and H). Furthermore, the stage-specific phosphoproteomes showed enrichment in functional networks linked to ribosomal biogenesis and RNA processing suggesting that protein kinases may control the establishment of stage-specific ribosomes. Finally, the identification of networks linked to proteasomal protein degradation in both ama and pro datasets not only further sustains the role of protein turnover in *Leishmania* stage differentiation, but uncovers a feedback loop with proteasomal activities controlling stage-specific degradation of protein kinases (see Figure 4), which in turn control stage-

545   specific ubiquitin ligases and deubiquitinating enzymes (see Figure 5D for phosphorylation pattern of

546   selected members of the Ubiquitin proteasomal system).

Discussion

Applying a 5-layer molecular systems analysis on spleen-derived bona fide amastigotes and culture-derived promastigotes, we uncovered a series of regulatory interactions during parasite stage differentiation that act within and between the levels of transcript stability, translational regulation, proteasomal protein turnover, and protein phosphorylation. Our data reveal an unexpected complexity of the *Leishmania* developmental process and raise important questions on the type of regulatory feedback loops that govern parasite stage differentiation, their biological integration, and their role in adaptation and phenotypic robustness.

A first series of regulatory networks identified in our experimental system rely on recursive or self-controlling interactions, where components of a given pathway are regulated by the pathway itself (Figure 6A). In promastigotes for example, a large regulon showing stage-specific increase in transcript abundance was defined by the GO term 'post-transcriptional regulation of gene expression'. In the absence of any significant gene dosage changes between amastigotes and promastigotes, and given the lack of transcriptional control in *Leishmania*, the differential expression of these transcripts itself must be regulated at the post-transcriptional level, thus establishing a recursive feedback loop. Our data uncovered a similar recursive feedback loop acting on the level of protein translation, which was revealed by a paradox: while amastigotes showed increased abundance for 38 transcripts annotated for the GO term 'ribosome biogenesis' (Table 4, sheet E), the corresponding proteins were of diminished abundance in amastigotes compared with promastigotes (Figure 3C). This dissociation between RNA and protein abundances indicates that ribosomal components themselves may be regulated at translational levels. Increased ribosomal protein abundance in promastigotes likely satisfies the requirement of increased translation capacity at this fast-growing stage (31, 93). In contrast, increased mRNA abundance of ribosomal components in amastigotes may indicate stage-specific mechanisms of mRNA storage (94) that may allow to jump start promastigote differentiation and accelerated growth following exposure to low temperature and neutral pH. This possibility is directly supported by our proteomics investigations, revealing ribosomal biogenesis as one of the first pathways induced over the initial 18h during the amastigote-to-promastigote developmental transition.

Differentiation to promastigotes was associated not only with a quantitative change of

ribosomes as indicated by the upregulation of a variety of ribosomal proteins and translation factors

(see Table 8), but also with a qualitative change of likely ribosome structure as indicated by the enriched

GO terms 'RNA methylation' and 'pseudouridine synthesis' that respectively include genes encoding

for 8 methyltransferases (e.g. the rRNA methyltransferases Ld1S_350672300 and Ld1S_350694000),

6 pseudouridine synthases (Table 8, sheets D to H), and the H/ACA snoRNP Nop10

(Ld1S_360722900). The differential expression of snoRNAs (see Figure 3D and Table 5) and the

developmentally regulated changes in rRNA modification that can affect ribosome activity (34) (Figure

3E and C) suggest the existence of stage-specific ribosomes. This finding is in line with our recent

reports providing indication for specialized ribosomes in *Leishmania* adaptation, in which we correlated

changes in rRNA pseudouridylation to *L. donovani* fitness gain in culture (31, 60).

Finally, a third recursive feedback loop acted at the post-translational level involving members

of various protein kinase families, which not only showed stage-specific expression, but also

represented a main class of phosphorylation substrates themselves. This recursive interaction likely

reflects stage-specific signaling cascades, where downstream kinases are activated through

phosphorylation by upstream kinases, such as exemplified by the MAP kinase pathway (92, 95). Indeed,

11 members of the MAP kinase family showed stage-specific changes in their phosphorylation pattern

(Figure 5D), 9 of which were only detected in the promastigote phospho-proteome and were previously

associated with this stage, being implicated in flagellar biogenesis (MPK9 and MPK3) or

metacyclogenesis (MPK4) (87, 91, 96). In contrast, the two MAP kinases MPK2 and MPK10 -

previously linked to amastigote differentiation and virulence (88, 91) - showed an amastigote-specific

increase in protein abundance and phospho-stoichiometry for defined phosphorylation sites.

Protein kinases not only engage in recursive regulatory interactions but are also at the center of

a second form of feedback loop defined by reciprocal interactions, where components of two or more

regulatory pathways can cross-control each other (Figure 6A). Phospho-stoichiometry analysis places

protein kinases at the center of three such loops: First, stage-specific kinase activities act on a series of

phosphatases that in return can modify kinase activities by dephosphorylation (Table 11, sheets D and

G). Second, 47 proteins associated with 'Translation' were identified in the stage-specific phospho-proteomes. Most of these proteins showed promastigote-specific phosphorylation, including 23 different ribosomal proteins and 15 eukaryotic translation initiation factors including eIF-2a, eIF-3 (subunits a, b, g, h), eIF-4A1, eIF-4G1-1, -2 and -5, or eIF-5a, of which many have been linked to various *Leishmania* biological process, such as translational control, cell cycle regulation or infectivity (29, 97-101). This opens the possibility of yet another reciprocal feedback loop, where phosphorylation of ribosomal components fine tunes translational control, which in turn may affect the expression of protein kinases themselves (102-110).

Finally, our results revealed a third reciprocal feedback loop between protein kinases and proteins linked to the proteasomal system. 11 ubiquitin hydrolases (also known as deubiquitinases, DUBs) and 10 ubiquitin ligases/transferases showed stage-specific phosphorylation at specific residues (Table 11, sheets D and I), while 14 protein kinases underwent differential proteasomal degradation (Figure 4F, Table 9, sheet G). These included two protein kinases that have previously been linked to *Leishmania in vivo* fitness (91), RDK1 (Repressor of differentiation kinase 1, Ld1S_320540600) and DIK (differentiation inhibitory kinase, Ld1S_280397500), whose rescue from degradation correlated with stalled amastigote-to-promastigote transition. Surprisingly, 5 protein kinases showed constitutive degradation across both stages. The constitutive expression of many proteasomal components across both amastigote and promastigote stages further supports a role of phosphorylation in regulating the observed, stage-specific proteasomal activities, considering that phosphorylation of DUBs has been linked to changes in their stability, localization, specificity or catabolic activity (111, 112).

Several arguments support a central role of this kinase/proteasome reciprocal feedback loop in early parasite development. First, proteasomal activities are known to allow for rapid proteome remodeling during trypanosomatid differentiation (81, 113, 114). Second, *Leishmania* stage differentiation is triggered by environmental pH and temperature changes, which can be sensed and translated into biological information by MAP kinase signaling cascades (84, 115). Finally, our data reveal a direct and essential role of this reciprocal regulation in the *Leishmania* amastigote-to-promastigote transition, which was abrogated in the presence of the proteasomal inhibitor lactacystin. This observation is reminiscent of previous reports genetically validating an essential role for the *L.*

mexicana ubiquitin conjugating (E2) enzymes UBC1/CDC34, UBC2 and UEV1, and the HECT E3 ubiquitin ligase HECT2 in promastigote to amastigote differentiation (81). Likewise, proteasomal activities were implicated in *T. cruzi* proliferation and differentiation (113, 116-118). The complexity of regulatory interactions between protein kinases and the proteasomal pathway is further increased by observations showing that substrate phosphorylation can mediate subsequent ubiquitination and degradation, while ubiquitination of MAP kinase family members can regulate kinase activity and localization rather than promote degradation (119-121). Future experiments simultaneously analyzing changes in both phosphoproteome and ubiquitome during *Leishmania* stage-differentiation may help elucidate this complex regulatory interaction and identify individual proteins and their PTMs that drive parasite development.

In conclusion, our data allow unprecedented insight into the regulatory landscape of *Leishmania* stage development. We reveal genetic feedback control as a central mechanism in parasite differentiation, with stage-specific interactions between genes and their products adapting the parasite phenotype to the environmental changes encountered inside its vertebrate and invertebrate hosts. In contrast to the major phenotypic shift observed between mammalian-stage amastigotes and insect-stage promastigotes, each developmental stage is stably maintained within its respective host even in the wake of environmental fluctuations. Such phenotypic robustness is known to rely on highly inter-connected, stage-specific regulatory networks (122), which in the case of *Leishmania* implicates recursive and reciprocal genetic interactions between protein translation, protein degradation and protein phosphorylation (Figure 6B and C). *Leishmania* differentiation thus represents an excellent model system to study how alternations between phenotypic plasticity (i.e. developmental transitions) and robustness (stable maintenance of stages) shape microbial fitness and allow for the emergence of adaptive biological traits. However, despite the robustness of a given stage, we previously showed that *Leishmania* still can fine-tune adaptive phenotypes in a given stage to promote fitness via interactions between gene dosage changes and compensatory responses at the post-transcriptional and translational levels (31, 60). This insight bears important consequences for future studies on *Leishmania* adaptation in clinical settings, which should combine multi-layer systems analyses such as described here with

660   deep learning algorithms that allow for data integration and the identification of novel biomarkers on

661   phenotypic and homeostatic rather than genetic levels.

 Figure legends

Figure 1: Comparative genomic analysis of *L. donovani* lesion-isolated amastigotes and derived

promastigotes. (A) Schematic overview of the parasite samples used for the various systems level

analyses presented in this study. A more detailed overview is shown in Figure S1. (B - D) Genomic

analyses. (B) Heatmap showing the somy score of three individual differentiation experiments using

hamster-isolated amastigotes (ama1, ama2 and ama3) and their corresponding, culture-derived

promastigotes analysed at passage 2 (pro1, pro2 and pro3). Samples and chromosomes are indicated on

the x- and y-axis, respectively. The somy scores were calculated as described in Material and Methods

and correlate to the grey level according to the shown legend. (C) Ratio plot showing the gene coverage

ratio ama versus pro (y-axis) for all genes across the 36 chromosomes (x-axis) as calculated based on

normalized median read depth. Each color represents one individual differentiation experiment using

ama isolates from individual hamsters. The red dotted lines indicate the ratio values corresponding to

1.5 (upper line) and 0.5 (lower line), indicating respectively gain or loss of one gene copy in the ama

samples. (D) Correlation plots representing the individual (histograms on the diagonal) and pairwise

(off-diagonal scatterplots) distributions of SNP frequencies for the indicated samples.

Figure 2: Stage-specific transcript profiling. (A) Pairwise scatter plots of the $\log_2$ (counts+1) for the

indicated samples. (B) Volcano plot (left panel) showing differential transcripts abundance between

two ama and pro samples as assessed by RNA-seq analysis. The dotted lines indicate fold change (FC)

= 2 (vertical line) and p-value = 0.01 (horizontal line). Transcripts with FC < 2 and adjusted p-value >

0.01 are represented by black dots. Transcripts with significant increased abundance FC ≥ 2 and

adjusted p-value < 0.01 in ama and pro are indicated respectively in dark cyan and dark grey. GO term

enrichment analysis of these transcripts for the category 'biological process' was performed, and

'cluster efficiency' (blue) and 'enrichment score' (orange) were plotted for the ama (middle panel) and

pro (right panel) datasets. The graphs show the main GO terms associated with a Benjamini and

Hochberg (BH) p-value < 0.05 that were identified using the BiNGO plug-in in Cytoscape. The color

intensity of each bar reflects the value of the BH p-value ranging from 0.05 (light blue or orange) to

less than 10^{-7} (dark blue or orange).

Figure 3: Systems analysis of the stage-specific transcriptome and proteome. (A) Volcano plot

showing differential protein abundance between four ama and pro samples as assessed by label-free,

quantitative proteomics analysis. The dotted lines indicate fold change (FC) = 2 (vertical line) and false

discovery rate (FDR) = 0.01 (horizontal line). Proteins with FC < 2 and FDR > 0.01 are represented by

light grey dots. Proteins that were reproducibly detected in only one of the two stages (considered

unique) are represented by the lateral histograms plotting their relative abundance according the iBAQ

value. These proteins together with proteins showing differential abundance of FC ≥ 2 and adjusted p-

value < 0.01 are indicated for ama in dark cyan and for pro in dark grey (see Table 6). GO term

enrichment analysis for the category 'biological process' was performed, and the resulting values for

'cluster efficiency' (blue) and 'enrichment score' (orange) were plotted for ama (middle panel) and pro

(right panel). The graphs show the main GO terms identified using the BiNGO plug-in in Cytoscape

and associated with a BH p-value < 0.05. The color intensity of each bar reflects the value of the BH p-

value ranging from 0.05 (light blue or orange) to less than 10^{-7} (dark blue or orange). (B) Double ratio

plot showing the log2 fold change between ama and pro in transcript (x-axis) and protein (y-axis)

abundances. For each dot the corresponding median absolute deviations calculated from protein and

transcript abundances are reported. Red dots correspond to changes in both transcript and protein

abundance with a p-value < 0.01. The dotted lines indicate fold changes (FC) = 2. The infinite symbol

(∞) represents proteins that are unique in the ama (upper signals) or pro (lower signals) datasets. Word

cloud enrichment performed with the *L. donovani* LdBPK orthologs is presented for each of the 4

quadrants, Q1 to Q4 and a detailed GO enrichment analysis is given in Table 7. The font size is

proportional to the number of genes per GO term and the grey scale refers to the p-value calculated for

each GO term. (C) KEGG map for ribosome biogenesis. Each gene associated with expression changes

indicated in Q1 to Q4 (see Figure 2C) and annotated with the GO term 'ribosome biogenesis' has been

projected on the ribosome biogenesis KEGG map. Each gene is represented by a square with the left

segment showing fold changes between ama and pro at transcript and the right segment showing fold

changes at the protein level, with the differential abundance indicated by the color and its intensity (red, increase; green, decrease). Only genes quantified in both transcriptome and proteome analyses and associated with an adjusted p-value < 0.01 were considered. (D) Volcano plot corresponding to the differential expression of snoRNA between pro and ama stages. Cyan dots (ama) and grey dots (pro) represent signals with FC ≥ 2 and adj. p-value < 0.05. Black dots represent non-significant changes. (E) Localization of differentially modified rRNA sites on the *Leishmania* ribosome. The RNA modifications are indicated as space filling in the *Leishmania* cryo-EM structure (PDB: 8RXH). The identity of the RNA modification is defined by the label (Ψ, pseudouridine; Gm and Cm refer to guanine and cytosine methylation, respectively; numbers indicate residues). Upregulated and downregulated RNA modifications in amastigotes are indicated in red and blue, respectively.

Figure 4: Analysis of stage-specific proteasomal protein turn-over. (A) Overview of experimental workflow. To assess early and late steps of differentiation (left part of the graph in black), comparative, quantitative proteomics analyses were performed on tissue-isolated amastigotes (ama), amastigotes at 18h into the transition to promastigotes in culture (ama-18h) and fully differentiated promastigotes after 1 passage *in vitro* (pro). To gain insight into stage-specific and constitutive protein degradation (right part of the graph in red), amastigotes and promastigotes were treated with the irreversible inhibitor lactacystin (ama-lacta and pro-lacta). (B) Microscopic images of amastigotes at 18h in absence of lactacystin treatment (amastigotes, left), after 48h in absence (-, middle) or presence of the inhibitor (+, right). (C) Western blot analysis. Protein extracts were obtained from 2×10^6 parasites after 3, 6, 18 and 48h of differentiation from amastigotes to promastigotes in presence (+) or absence (-) of lactacystin. Extracts were separated by electrophoresis, transferred onto PVDF membrane and the presence of PFR2 was revealed using an anti-PFR2 antibody (upper panel). A Coomassie blue stain of the same gel is shown as loading control (lower panel). (D) Double ratio plot comparing the log2 fold changes in protein abundance between ama-18h/ama-lacta (x-axis) and ama/ama-lacta (y-axis). Dashed lines indicate FC = 2 and the infinite symbol (∞) represent proteins that are unique to one or the other comparison. (E) Projection of the proteins stabilized in presence of the inhibitor in ama (ama-lacta, orange) and pro (pro-lacta, pink) onto the double ratio plot shown in Figure 3B. Note that ama-specific

proteins are rescued from degradation in pro-lacta, while pro-specific proteins are rescued from

degradation in ama-lacta. (F) List of the protein kinases stabilized after lactacystin treatment in pro or

ama (columns lacta rescued) and either not detected in the total proteome (column ama vs pro, grey

cell) or specifically quantified at the ama (red cell) or the pro (green cell) stage. (G) Venn diagram

showing the number of proteins specifically stabilized in ama or pro or stabilized in both stages after

lactacystin treatment.

Figure 5: Stage-specific phospho-proteomic profiling. (A) Volcano plots corresponding to the total

proteome (right panel) and phosphoproteome (left panel) analyses. Proteins and phosphosites (P-sites)

only identified in one stage are presented at each side of the volcano plots. Cyan dots (ama) and grey

dots (pro) represent signals with FC $\geq$ 2 and FDR < 1%. Light grey dots represent non-significant

changes. (B) Phosphostochiometry. Ratio plot comparing the log2 fold changes in total protein

abundance between ama and pro (x-axis) and phosphosite abundance in ama versus pro (y-axis).

Dashed lines indicate a FC = 2. The color intensity of each dot reflects the p-value calculated for the

phosphostochiometry analysis as indicated in the graph. (C) GO term enrichment analysis for the

category 'biological process. Only stage-specific phosphosites were considered (i.e. sites that showed

a significant increase in phosphostochiometry, and sites that were only detected in one or the other

stage, whether the protein was identified or not in the total proteome analysis). The histogram plots

show 'cluster efficiency' in blue and 'enrichment score' in orange for the GO term enrichment analysis

performed with the ama (middle panel) and pro (right panel) datasets. The main GO terms identified

using the BiNGO plug-in in Cytoscape and associated with a BH p-value < 0.05 are shown. The color

intensity of each bar reflects the value of the BH p-value ranging from 0.05 (light blue or orange) to

less than 10^{-7} (dark blue or orange). (D) Phosphorylation pattern of MAPK and MAPKK proteins (left

panel) or Ubiquitin hydrolases and transferases (right panel) identified in the phosphostochiometry

analysis. The positions of the phosphorylated amino acid specific to ama (red) and pro (green) are

indicated. Only phosphopeptides identified in the phosphostochiometry analysis (see Table 11) and

detected either in one stage or the other or quantified with a FC $\geq$ 2 were considered for the identification

of the phosphosites.

Figure 6: Network analyses. (A) Model of *Leishmania* gene expression regulation during stage differentiation. (B-C) Network analysis. Networks, restricted to the protein kinases (blue nodes), proteins implicated in ribosome biogenesis and ribosomal proteins (violet nodes) and proteins linked to ubiquitination/deubiquitination (orange nodes) identified from the phosphostochiometry analysis conducted in ama (B) or pro (C), were generated with the STRING plug-in of the Cytoscape software package using *L. infantum* orthologs. Each node represents a phosphoprotein with the respective gene identifier indicated. GO terms associated to the proteins are represented by the colored segments around the nodes according to the legend shown in the graph. Only GO terms associated with a p-value < 0.05 were considered.

Sup figure legends

Figure S1: Overview of the samples used in this study. Each *L. donovani* infected hamster is identified by the cage number, as are the hamster-derived amastigotes (ama) and corresponding promastigotes (pro). Only samples indicated by boxes were used for the various analyses as indicated.

Figure S2: (A) Box plots representing the median somy score for all the chromosomes. Each box corresponds to one of three biological replicates for ama and pro as indicated in the graph. Sample identifiers are detailed in the legend of Figure 1A and Figure S1. (B) Frequency distribution plots representing, for each biological replicate and each chromosome, a kernel density estimate of the distribution of SNP frequencies in the amastigote and promastigote sample.

Figure S3: (A) Cluster analysis of differentially expressed genes observed from RNA sequencing analysis. The heatmap is based on the variance-stabilized transformed count matrix. Rows and columns have been re-ordered thanks to a hierarchical clustering (using the correlation and Euclidean distances respectively and the Ward aggregation criterion). Color scale ranges from -1.5 to +1.5 as the rows of the matrix have been centered.

Figure S4: (A) Cluster analysis of metabolites showing differential abundance in ama and pro samples. (B) Quantification of amino-acids or metabolites involved in glycolysis, TCA cycle and mannogen cycle. The y-axis represents the intensity for each metabolite as detected by MS. (C-D) Correlation between metabolomic and proteomic analyses for TCA cycle (C) and Glycolysis/Gluconeogenesis (D). Fold changes in proteins abundance (squares) and metabolite abundance (circles) with adjusted p-values of respectively < 0.01 and < 0.05 were projected on the respective KEGG maps. The color intensity reflects the fold change value as indicated by the legends in the graphs. (E) Representation of the mannogen pathway. Metabolites are represented by a square, with red-lined squares corresponding to the metabolites that accumulate at the ama stage. Proteins are represented by a circle, with fold changes in protein abundance between ama and pro datasets indicated by color and intensity (red, increase; green, decrease).

Figure S5: Label free quantitative proteomic analyses in presence or absence of lactacystin. (A) to (E) Volcano plots representing changes in protein abundance for the different comparisons (left panels) and GO term enrichment analyses for the indicated sample (right panels). Proteins identified by at least two peptides in at least three out of four biological replicates were considered. For volcano plots, colored dots indicate values with FDR < 0.01 and FC ≥ 2 (see Table 8). The light grey dots indicate non-significant expression changes. The colored bars indicate unique protein identifications in one (green) or the other (orange) condition with relative abundance indicated by the iBAQ value (left panels). Proteins quantified in only one condition or showing a significant increase in abundance (FC ≥ 2 and adjusted p-value < 0.01) were used to perform the GO analysis for the category 'biological process'. The histogram plots show 'cluster efficiency' in blue and 'enrichment score' in orange (right panels). (A) ama vs ama-18h; (B) ama-18h vs pro; (C) ama vs ama lacta; (D) ama-18h vs ama lacta; (E) pro vs pro lacta.

Figure S6: (A) Functional network enrichment analysis for proteins showing increased abundance in ama-18h compared to ama. The network was generated with the *L. infantum* orthologs using the STRING plug-in of the Cytoscape software package. The GO terms associated with each protein are

represented by the colored segments according to the legend. Only GO terms associated with a p-value

< 0.05 were considered. (B) KEGG map for ribosome biogenesis showing changes in abundance protein

between ama and ama-18h. Only proteins with an adjusted p-value < 0.01 were considered. Differential

abundance is indicated by the color and its intensity (red, increase; green, decrease).

Figure S7: (A) Viability assessment of the indicated amastigote (left panel) and promastigote (right

panel) samples after lactacystin treatment by FACS analysis using propidium iodide or YoPro staining.

(B) Micrographs showing hamster-derived amastigotes during *in vitro* differentiation to promastigotes.

Parasites were collected at the time points indicated, fixed in 4% PFA, and seeded on poly-L-lysine-

treated coverslips. Parasites were observed with a 63x oil immersion objective and images were

processed with AxioVision Rel.4.8 software. Differential interference contrast (DIC) microscopy was

performed with Axioplan 2 imaging microscope using a 63x oil immersion objective, Axiovision

software and AxioCam MRm camera (Carl Zeiss). The scale bars correspond to 1 μm. (C) Western blot

analysis of protein extracts obtained at the indicated time points after *in vitro* differentiation from

hamster-derived amastigotes to promastigotes. 10 μg of protein were separated on NuPAGE 4-12%,

transferred onto PVDF membrane subsequently incubated with antibodies against paraflagellar rod

proteins 1 and 2 and alpha-tubulin, and revealed with HRP-conjugated secondary antibodies.

Figure S8: Functional network enrichment analysis for the proteins stabilized in presence of lactacystin

in promastigote (pro) and amastigote (ama) datasets. The network was generated with the *L. infantum*

orthologs using the STRING plug-in of the Cytoscape software package. The GO terms associated with

each protein are represented by the coloured segments according to the legend. Only GO terms

associated with a p-value < 0.05 were considered.

Figure S9: List of protein kinases stabilized in presence of lactacystin. Their corresponding orthologs

in *L. donovani* strain LdBPK and *L. mexicana* are indicated, as well as the knock-out phenotype as

published by Baker et al (2021).

Figure S10: Upset plot representing the number of phosphopeptides (y-axis, blue histograms, numbers

are indicated) and the samples in which they were quantified (x-axis). Pie chart of the phosphopeptide

analysis showing the distribution of the phosphorylation sites according to the condition in which they

were quantified as indicated in the legend of the graph.

Figure S11: Functional Network analysis. Networks for proteins that show unique or increased

phosphorylation in ama (upper panel) and in pro (lower panel) datasets were generated with the

STRING plug-in of the Cytoscape software package using *L. infantum* orthologs. The nodes represent

phosphoproteins associated with one or more GO terms (listed in the graph), with the respective gene

identifier indicated. The GO terms associated to the proteins are represented by the colored segments

around the nodes according to the legend shown in the graph. Only GO terms associated with a p-value

< 0.05 were considered.

Acknowledgments

We thank the CEA-CNRGH for its contribution to the sequencing costs and all the CEA-CNRGH
staff who performed sample preparation and sequencing for their excellent technical assistance.

We acknowledge the help of the HPC Core Facility of the Institut Pasteur for this work.

We thank Karim Sébastien from the animal facility of the Institut Pasteur for his commitment in the
care of our animals.

This work was supported by the Agence Nationale pour la Recherche Labex 'Integrative Biology of
Emerging Infectious Diseases' contract ANR-10-LABX-62-IBEID and Labex 'French Alliance for
Parasitology and Health Care' contract ANR-11-LABX-0024 (GFS, PP), the France Génomique
National infrastructure, funded as part of the « Investissements d'Avenir » program managed by the
Agence Nationale pour la Recherche contract ANR-10-INBS-09 (HV, RL, CP), and the ERC
SYNERGY project DecoLeishRN, Grant agreement ID: 101071613.

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1345

Figure 1

A

B

Figure 2

Figure 3

A

B

amastigotes | (-) lacta | (+) lacta

C

D

E

F

Gene ID	Annotation	ama vs pro	lacta rescued	
			ama	pro
Ld1S_280397500	differentiation inhibitory kinase	ama-specifc detection		X
Ld1S_320543000	protein kinase			X
Ld1S_340649100	protein kinase			X
Ld1S_350674800	casein kinase I			X
Ld1S_050815400	protein kinase	pro-specifc detection	X	
Ld1S_120047800	cell division protein kinase		X	
Ld1S_150105800	protein kinase		X	
Ld1S_320540600	repressor of differentiation kinase 1			X
Ld1S_190177600	protein kinase domain containing protein			X
Ld1S_040809500	casein kinase I	not detected		X
Ld1S_080866500	protein kinase			X
Ld1S_180153200	ser/thr kinase			X
Ld1S_210228200	protein kinase		X	
Ld1S_220237000	protein kinase		X	X

G

Lactacystin-stabilized in:

117 ama | 12 both | 277 pro

Figure 4

Figure 5

Figure 6

Supp Figures

Sup Figure S1

Sup Figure S2

A

ama 4   ama 5   pro 4   pro 5

Sup Figure S3

Sup Figure S4

Sup Figure S5

Sup Figure S6

Sup Figure S7

Sup Figure S8

ID Ld1S	ID LdBPK	ID LmxM	Annotation	KO phenotype*
Ld1S_320540600	LdBPK_320860.1	LmxM.31.0810.1	Repressor of differentiation kinase 1, RDK1	Loss of fitness during pro to ama differentiation Loss of fitness in animal infection
Ld1S_220244300	LdBPK_220970.1	LmxM.22.1150.1	Protein kinase domain containing protein, putative	
Ld1S_190177600	LdBPK_191290.1	LmxM.19.1300	hypothetical protein, unknown function (PP: Protein kinase domain containing protein, putative)	
Ld1S_180153200	LdBPK_180650.1	LmxM.07.0900	serine/threonine kinase-like protein, putative	
Ld1S_080866500	LdBPK_080870.1	LmxM.08.0930.1	protein kinase, putative	
Ld1S_040809500	LdBPK_041230.1	LmxM.04.1210.1	casein kinase I, putative	
Ld1S_350674800	LdBPK_351020.1	LmxM.34.1000.1	casein kinase I, putative	
Ld1S_340649100	LdBPK_342850.1	LmxM.33.3020.1	protein kinase, putative	Loss of fitness in animal infection
Ld1S_320543000	LdBPK_321080.1	LmxM.31.1020.1	protein kinase, putative	
Ld1S_280397500	LdBPK_281800.1	LmxM.28.1670.1	differentiation inhibitory kinase, putative, ZFK	Moderate loss of fitness in animal infection
Ld1S_220237000	LdBPK_220370.1	LmxM.22.0490.1	protein kinase, putative	ESSENTIAL
Ld1S_210228200	LdBPK_210330.1	LmxM.21.0270.1	protein kinase, putative	Loss of fitness in animal infection
Ld1S_150105800	LdBPK_151180.1	LmxM.15.1200	protein kinase, putative	
Ld1S_120047800	LdBPK_120130.1	LmxM.12.0130.1	Cell division protein kinase, putative	
Ld1S_050815400	LdBPK_050550.1	LmxM.05.0550.1	protein kinase, putative	

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ID Ld1S	ID LdBPK	ID LmxM	Annotation	KO phenotype*

Sup Figure S9

Sup Figure S11