Cell-Autonomous and Systemic Circadian Regulation of Gene Expression in Adipocytes

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Running title: Circadian Gene Regulation in Adipocytes

ABSTRACT

Circadian clocks strongly influence adipocyte biology. Studying adipocyte circadian regulation in the

absence of organismal cues isolates cell-autonomous clock control, providing insight into mechanisms

relevant to metabolic disease. To resolve circadian core biological programs, we deeply sequenced bulk

RNA from inguinal-derived, *in vitro*-differentiated adipocytes (IVDAs) over 2.5 days and compared

clock-controlled genes (CCGs) with existing mouse supraclavicular brown (BAT) and epididymal white

(eWAT) adipose tissue circadian time-course datasets. Using Phase Set Enrichment Analysis (PSEA), we

report 21% of the protein-coding transcriptome is rhythmic in IVDAs. Intrinsic circadian regulation

governs key processes in energy metabolism, molecular transport and transcription. Integration with *in*

vivo datasets reveals that BAT and IVDAs exhibit more cohesive rhythmic pathways than does eWAT,

clustering around the late-night early-morning transition. To explore how these pathways may be

regulated, we reexamined a recent interscapular BAT cistrome dataset. Using IVDA transcription factors

that were phase-aligned (≤4hr) with *in vivo* as input, motif enrichment analysis revealed two temporally

distinct regulatory programs; an early E-box activator ARNT-family/bHLH-PAS program was enriched

for transcriptional regulation, RNA metabolism, and signaling pathways, and a late nuclear receptor-

associated program enriched for energy metabolism, phospholipid biosynthesis, mitochondrial function,

ECM organization, and nuclear receptor signaling. Overall, we identify novel rhythmic transcripts and

define cell-autonomous circadian programs in adipocytes whose timing is further sculpted by systemic

cues *in vivo*. Because obesity is associated with adipocyte hypertrophy and hyperplasia, processes likely

influenced by circadian regulation, these findings advance our understanding of clock-controlled

adipocyte metabolism and its contribution to metabolic dysfunction.

INTRODUCTION

The circadian clock plays a central role in regulating metabolism by synchronizing physiological

processes to the 24h daily cycle (Bass, 2012; Hurley et al., 2016; Schrader et al., 2024). In nearly every

animal cell, this internal timekeeper coordinates key functions such as energy expenditure and

intermediary metabolism (van Rosmalen et al., 2024; Neufeld-Cohen et al., 2016; Malik et al., 2024).

While studies of adipose tissue *in vivo* have established the rhythmic nature of adipose tissue function, the

extent to which these rhythms are cell-intrinsic versus systemically driven remains unclear. Adipocytes

integrate nutrient availability, hormonal signals, and energetic demand and retain the capacity for

autonomous circadian regulation. Clocks in adipose tissues drive rhythmic expression of genes and

proteins critical to nutrient metabolism, signaling, and secretion, which influence glucose homeostasis

and energy balance (Zvonic et al., 2006; Rosen and Spiegelman, 2006; Hepler and Bass, 2023). Adipose

tissue is composed of several cell types, including mesenchymal stem cells, immune cells and vascular

cells, with distinct anatomical depots serving specialized functions (Sakers et al., 2022; Yang Loureiro et

al., 2022; Corvera et al., 2026). In addition, recent single-cell and single-nucleus studies have

demonstrated substantial heterogeneity among adipocytes within individual depots, revealing multiple

transcriptional states and functional subpopulations (Loft et al., 2025). While all white adipose tissue

(WAT) stores energy as triglycerides in lipid droplets, brown adipose tissue (BAT) is also dense with

mitochondria and constitutively capable of dissipating chemical energy as heat by adaptive non-shivering

thermogenesis (ANST) (Bunk et al., 2025; Rahbani et al., 2024; Sun et al., 2021; Cannon and

Nedergaard, 2004; Puigserver et al., 1998). Interspersed throughout inguinal WAT (iWAT) and other

depots is a subtype transiently capable of ANST, referred to as "brown in white" or beige (Vargas-Castillo

et al., 2024; Bertholet et al., 2017; reviewed in Inagaki et al., 2016). Over-feeding or cold exposure via

sympathetic nervous system activation induces a metabolic sink via H^+ leakage across the mitochondrial

membrane or futile phosphorylation cycling, and this is retained through epigenomic memory (Roh et al.,

2018). While the physiological relevance and mechanisms underlying this plasticity are debated (Cypess

et al., 2025), the upregulation of an ANST-like response in adipocytes is, in part, driven by cell-

autonomous sensing of ambient cold (Ye et al., 2013).

The circadian oscillator comprises a self-regulating transcriptional-translational feedback loop (TTFL).

The core clock proteins CLOCK and BMAL1 form a heterodimer that enters the nucleus and binds to E-

box sequences on target clock-controlled genes (CCGs) to initiate transcription (Takahashi, 2017; Van

Drunen and Eckel-Mahan, 2021; Rasmussen et al., 2022). Upregulated genes include paralogs of

Cryptochrome (*Cry*) and Period (*Per*), whose products form multimeric complexes with casein kinase

1d/e; these interact with CLOCK:BMAL1, resulting in their inhibition by blocking the activity of

CLOCK and BMAL1 at E-boxes and thereby also reducing *Per/Cry* expression (Smyllie et al., 2025).

Serial phosphorylation of the CRY and PER proteins by CK1 eventually relieves inhibition, allowing the

cycle to reinitiate *Cry* and *Per* expression (Cao et al., 2021; Narasimamurthy and Virshup, 2021; Philpott

et al., 2023; Ricci et al., 2025).

Disruption of circadian rhythms is strongly associated with metabolic dysfunction, including obesity and

insulin resistance. When food or physical activity, very strong temporal cues, are shifted out of the active

phase and into the resting phase, body composition as well as other parameters of metabolic health are

altered (Acosta-Rodríguez et al., 2022; Acosta-Rodríguez et al., 2024), yet it remains unclear which

aspects of adipocyte metabolic regulation arise from intrinsic clock control versus systemic cues

including feeding, temperature, and neuroendocrine input (Allada and Bass, 2021; Fekry and Eckel-

Mahan, 2022, Schrader et al., 2024). Resolving this distinction is essential for understanding how

temporal regulation of adipocyte metabolism contributes to health and disease (Deota et al., 2025).

Indeed, circadian regulation in adipose tissue has significant implications. In mice, chronic circadian

shifts drive WAT hypertrophy, inflammation, and disrupted metabolic regulation, resulting in obesity-like

transcriptional changes and impaired insulin signaling (Xiong et al., 2021). Circadian disruptors like shift

work and jet lag alter adipokine expression and secretion, linking disrupted rhythms to metabolic

dysfunction (Kettner et al., 2015; Scheer et al., 2009; Kiehn et al., 2017; Samanta and Ali, 2022). Brown

adipose-specific deletion of *Arntl/Bmal1* disrupts thermogenesis and energy expenditure and increases

susceptibility to developing obesity when mice are fed a high fat diet (Hasan et al., 2021). Characterizing

the circadian regulation of target pathways offers insight into the pathogenesis and treatment of metabolic

syndrome.

Here, we isolate adipocyte circadian biology from organismal context by comparing circadian

transcriptomes of *in vitro* differentiated adipocytes (IVDAs) with *in vivo* brown and white adipose depots.

This study used developmentally equivalent cells but excludes input from other cell types and from

systemic influences such as feeding, hormones, or body temperature. Accordingly, our experimental

design is optimized to capture cell-intrinsic circadian programs in differentiated adipocytes. Our

differentiation protocol pushes cells into a committed adipocyte state and does not attempt to capture the

adipocyte subtype heterogeneity revealed by single-cell profiling *in vivo* (Maniyadath et al., 2023).

Through the systematic comparison of our deeply sequenced transcriptome with comparable datasets

derived from whole animals (Zhang et al., 2014), we find that the cell-intrinsic circadian clock regulates

the gene expression of fuel oxidation, adipose remodeling, and intra-organ communication programs. We

have focused on identifying which pathways rely on systemic signals for their timing as well as the ways

in which adipocytes execute these elegant, interlocking pathways. Our findings support the robust

circadian timing of mitochondrial oxidative metabolism, energy metabolism, extracellular matrix (ECM)

homeostasis, molecular transport, and transcriptional regulation, while phase alignment is selectively

shifted by tissue-specific systemic cues. Together, these data demonstrate that adipocyte circadian output

is controlled by cell-intrinsic temporal regulatory mechanisms, while tissue-specific systemic cues

modulate phase alignment rather than generate rhythmicity *de novo*.

RESULTS

The transcriptome of differentiated adipocytes in culture shows widespread circadian regulation.

We isolated pre-adipocytes of inguinal white adipose tissue (iWAT) obtained from Per2::Luc C57BL/6J

mice. The pre-adipocytes were then differentiated *in vitro*, and their clocks synchronized via serum shock

and forskolin treatment (see Methods and Supplemental Methods for further details). Synchronized

Per2::Luc adipocytes exhibited robust ~24h PER2 oscillations, which were confirmed at the mRNA level

by qRT-PCR (Fig. 1A, Fig. S1A-D). Sampling for two replicate time series began at 12 hours post-

synchronization (HPS) and continued until 72 HPS. RNA-seq analysis yielded >60M paired-end reads per

sample, a depth we chose to detect low-abundance transcripts. To our knowledge, this dataset represents

the longest, most deeply sequenced, highest resolution circadian transcriptome for differentiated

adipocytes in cell culture to date and has allowed us to detect previously undetected rhythms in CCGs.

We performed analysis at the transcript rather than the gene level for protein-coding genes (Fig. S2,

Methods, Supplemental Methods). In this dataset, core clock genes were highly circadian, and replicates

were almost uniformly in phase (Fig. 1B and Fig. S3). Using ECHO (De Los Santos et al., 2020) and

MetaCycle (Wu et al., 2016) with a Benjamini-Hochberg (BH) adjusted p-value ≤ 0.005 and 22-30h

period, we identified 4,920 unique circadian transcripts in IVDAs corresponding to 4,346 unique CCGs

(21% of the coding transcriptome) (Fig. 1C, D, Supplemental File 1). This represents approximately three

times as many CCGs as previously reported for BAT and five as many as was reported for eWAT (cf.

Zhang et al., 2014), reflecting use of RNA-seq vs. microarrays, transcript vs. gene level analysis, and

ECHO and MetaCycle vs. JTK cycle. ECHO period estimates for rhythmic transcripts in IVDAs are fairly

uniform across the detected range (Fig. S4A), the majority of circadian transcripts had BH adj p-values ≤

0.005, indicating accurate statistical modeling of these genes by ECHO (Fig. S4B), and the distribution of

oscillation types shows approximately equal proportions of harmonic and damped oscillations (Fig. S4C)

(De Los Santos et al., 2020). As an independent check of our analysis pipeline, as well as to estimate how

many CCGs were primary targets of the core clock, we interrogated our list of CCGs against the 6,276

unique BMAL1 target genes identified by iWAT BMAL1 ChIP-seq (Hepler et al., 2022; GEO accession #

GSE181443) defining BMAL1-associated targets using a −10 kb to +1 kb window relative to the

transcription start site (TSS). The percentage of these CCGs identified as BMAL1 targets does not change

significantly as we increase stringency, providing additional validation that our list of CCGs are indeed

clock-controlled (Table S1).

Comparison of the *in vitro* circadian transcriptome with *in vivo* data from BAT and eWAT.

A principal goal of this study is to ascertain the relative importance of cell-autonomous versus systemic

cues in the rhythmicity of adipocytes. To understand how CCGs identified in our study compare to

circadian transcriptomes of mouse adipose tissue influenced by systemic cues, we obtained *in vivo* BAT

and eWAT (epididymal white adipose tissue) circadian transcriptome data generated using microarrays

(Zhang et al., 2014, GEO accession # GSE54652). We filtered genes to remove transcriptional noise as

was done for IVDAs, resulting in a detected protein-coding transcriptome of 11,249 genes for BAT and

11,624 for eWAT (see Methods, Supplemental Files 2 and 3). To ensure accurate comparisons with our

dataset, we reanalyzed these datasets using our analysis. ECHO and MetaCycle identified 4,061 rhythmic

protein-coding CCGs in BAT and 3,157 in eWAT. This represents 19.3% for BAT and 15% for eWAT,

similar to that calculated for IVDAs (Fig. 1C, D) but higher than previously reported (BAT = 8% and

eWAT = 4% of protein-coding genes), reflecting the use of ECHO as well as our application of a wider

period range compared to Zhang et al. Circadian time enables us to use a common axis to compare phase

between organisms with different free-running period lengths. *Per2* expression is robustly circadian in

every rhythmic mammalian tissue and consistently peaks about 14h after a dark-to-light transition in

adipose tissue from mice (Zhang et al., 2014). To assign an estimate of circadian time (CT) to IVDAs, we

used the peak of *Per2* expression (22 HPS) following release from synchronization as a phase reference

point, assigning this a phase of CT14 and calculating phases for other genes in reference to *Per2*. By this

metric, phases of rhythmic core clock component genes or their paralogs in our IVDAs (e.g., *Npas2* and

Clock) are within 2 hours of those reported from BAT and eWAT *in vivo* (Zhang et al., 2014) (Table S2,

Fig. S5). These results demonstrate that the cell-intrinsic TTFL is functioning similarly in IVDAs when

compared to adipose tissue, such that observed differences in the phasing of circadian transcripts are not

due to differences in the core oscillator, but rather in the circadian physiology of the IVDAs vs. *in vivo*-

sampled adipose.

Cell-autonomous circadian regulation of a breadth of processes in adipocytes is frequently

further sculpted *in vivo* by cell-extrinsic factors.

To identify which biological pathways were circadianly regulated and determine whether their timing is

dependent on systemic cues, we combined standard enrichment analysis with Phase Set Enrichment

Analysis (PSEA), both of which utilized Reactome gene sets (Zhang et al., 2016; Milacic et al., 2024)

(Fig. 2). We first conducted a Reactome pathway enrichment analysis of all IVDA CCGs using

PANTHER (Thomas et al., 2022; Ashburner et al., 2000; The Gene Ontology Consortium, 2023) and

identified RHO, CDC42, RAC, and MIRO GTPase pathways, along with membrane trafficking,

transcription, RNA and protein metabolism, cell cycle, and signaling pathways, as significantly enriched

for circadian genes (FDR ≤ 0.05) (Fig. 2A). Phase-clustered gene sets are defined as gene sets whose

expression peaks occur close together in time. Though PANTHER identifies pathways enriched for

circadian genes, it does so without regard to peak phase. To determine whether circadian genes in IVDAs,

BAT, and eWAT were phase-clustered, we applied Phase Set Enrichment Analysis (PSEA; Zhang et al.,

2016). PSEA Kuiper scores quantify the degree to which circadian gene CT values deviate from a

uniform distribution, and pathways with Kuiper q ≤ 0.1 were considered significantly phase-clustered

(Methods). Phase clustering strength was further assessed using the vector-average magnitude metric,

which reflects temporal cohesiveness within each gene set.

For easier interpretation and comparison between *in vitro* and *in vivo* datasets we grouped

phase-clustered pathways into functional categories using top-level Reactome terms and visualizing the

distribution of peak CTs across tissues (Fig. 2B). Although many pathways are shared between IVDAs,

BAT, and eWAT, their circadian timing differs systematically by tissue. Most pathways, including those

relating to biological process, immune system, protein metabolism and post-translational modifications

(PTMs), cell cycle, signal transduction, and transcription generally peak later in IVDAs than *in vivo*

depots, indicating widespread temporal repositioning of circadian programs in IVDAs compared to those

in vivo (Fig. 2). In contrast, the timing of the energy metabolism pathways was more closely aligned

between IVDAs and BAT, while eWAT was selectively shifted. ECM homeostasis pathways exhibit

conserved timing across all three tissues, clustering near the day-night transition. The timing of molecular

transport pathways was coordinated in IVDAs and BAT with no detected phase-clustering in eWAT.

Together, these patterns show that circadian pathway timing depends on both functional category and

tissue context, with some programs maintaining similar phase relationships.

To ensure phase clustering isn't driven by the repeated detection of the same circadian genes across

related Reactome terms, we performed a PSEA leading-edge overlap analysis. Pairwise Jaccard similarity

among the top 50 most overlapping pathways reveals high gene sharing only among cell-cycle-related

pathways; most other functional categories show low overlap (Fig. S6). Phase clustering across diverse

biological processes arises from independent circadian gene modules rather than a single dominant driver

set, consistent with the high overlap between pathways strongly phase clustered and those highly enriched

for circadian genes by PANTHER (Fig. S7). Comparison of vector-average magnitude shows no

difference among phase-clustered pathways across all tissues (Fig. S8), indicating that systemic cues

primarily modulate pathway timing rather than altering overall temporal cohesiveness.

To measure the degree of cell-autonomous versus systemic circadian regulation of phase-clustered

pathways, we compared those detected in IVDAs, BAT, and eWAT and found a high overlap (101

pathways in both BAT and IVDAs, 35 in both eWAT and IVDAs, and 33 shared across all three datasets)

(Fig. 2C, top and Fig. S10). A higher resolution representation of all phase-clustered pathways further

highlights conservation of circadian pathway architecture with depot-specific phase shifts driven by

systemic cues (Fig. S11). BAT-IVDA pathways exhibit a consistent average phase delay of about 8-9

hours, whereas no dominant phase-shift pattern was observed for eWAT (Fig. 2C, bottom and Fig. S11).

Pathways conserved and in phase across all three depots were dominated by ECM homeostasis, immune

signaling, and receptor tyrosine kinase (RTK) pathways (Fig. 2D and Fig. S10). In contrast, 17 pathways

were rhythmic in both BAT and eWAT but not in IVDAs, indicating circadian regulation that depends on

systemic inputs rather than adipocyte-intrinsic oscillators (Fig. S12). These pathways were enriched for

ECM organization, immune and growth factor signaling, and cholesterol biosynthesis, and peaked during

late night to early morning (CT20-4) in both depots. Taken together, these results demonstrate that

circadian regulation of adipocyte biology is neither purely intrinsic nor purely systemic but rather

emerges from the integration of autonomous oscillators with depot-specific physiological cues that sculpt

pathway timing *in vivo*.

Clock-organized biological work includes energy metabolism.

Metabolic processes follow a daily rhythm such that anabolic processes (e.g., glycolysis and lipogenesis)

primarily occur during the feeding phase and catabolic processes including lipolysis and fatty acid

oxidation (FAO) are engaged during fasting (Liu et al., 2025; Lekkas and Paschos, 2019). Reactome

metabolic pathways are strongly enriched for CCGs, with pronounced temporal organization in IVDAs

(Fig. 3A, B). Lipid-associated pathways are phase-aligned between BAT and IVDAs but diverge in eWAT,

indicating depot-specific regulation of adipocyte metabolism (Fig. 3B and Figs. S9-11).

The top-level Reactome metabolism pathway shows a significant Kuiper statistic (6×10^{-4}) and a very low

magnitude (0.14), indicating a bimodal distribution of metabolic CCGs (Fig. 3B, Supplemental File 5).

Phase distribution of these genes reveals two dominant peaks: a subjective-day cluster (CT5-16) and a

subjective-night cluster (CT17-4) (Fig. 3C). To identify which metabolic processes were specific to each

peak, we performed PSEA on the CT5-16 and CT17-4 CCGs independently using only the gene sets

which fall under the top-level term of Metabolism. This analysis revealed a clear separation of opposing

metabolic processes.

In mice, which are nocturnal, the subjective day overlaps with the resting period (dawn to dusk) whereas

the subjective night overlaps with the active period (dusk to dawn). Pathways phase-clustered exclusively

within the CT5-16 window are associated with anabolic and biosynthetic processes, consistent with

processes happening during the rest-phase. Pathways phase-clustered exclusively within the CT17-4

window are enriched for lipid-centric pathways, including mitochondrial and peroxisomal FAO,

phospholipid metabolism, and lipid remodeling which are expected to dominate during the active phase

(Hepler et al., 2022) (Fig. 3D). Despite the two windows containing a similar number of CCGs, the

nighttime window yielded far more phase-clustered pathways specific to that window suggesting it is the

major metabolic window in IVDAs. The CCGs in these pathways were sharply in phase and no genes

occur in more than 3 pathways (data not shown), consistent with these being distinct circadianly regulated

programs.

Multiple rate-limiting enzymes within these pathways are circadian in IVDAs and in phase with BAT.

Enzymes in glycolysis (*Pfkm, Pfkp*) peak late in the circadian cycle (IVDA CT20, 24; BAT CT20, 2).

Acaca, which catalyzes the first irreversible step in fatty acid synthesis, is circadian and in phase between

IVDA and BAT peaking at CT23 and CT20, respectively. Lipolytic enzymes *Pnpla2 and Lipe* peak at

CT2 in IVDAs and CT24 in BAT which coincides with the phase of triglyceride mobilization. Fatty acid

oxidation is rhythmically controlled by *Cpt1a, Cpt1b, Acadl, Hadha, Hadhb*, and *Acox1* (Liu et al, 2025)

which are tightly phase-clustered in the very early morning (IVDA CT2-4; BAT CT2-6). Oxidative

metabolism through the TCA cycle (*Ogdh, Idh2, Mdh1*) and branched-chain amino acid degradation

(*Bckdha*) also peaks in this window. Additional rate-limiting steps in phospholipid synthesis (*Gpat3*) and

NAD$^+$ salvage (*Nampt*) peak in phase with lipid handling and mitochondrial function (Supplemental File

1 and 2). Together, these data suggest circadian timing preferentially constrains pathways that support

sustained oxidative metabolism, with rhythmic pathways sharing similar timing persisting in cultured

adipocytes without systemic cues, underscoring their cell-autonomous rhythmicity.

Pathways associated with the ECM, including integrin and focal adhesion signaling, ECM-receptor

interactions, and heparan sulfate and glycosaminoglycan metabolism, show strong phase clustering and

peak during the late active to early rest window across IVDAs, BAT, and eWAT (Fig. 2D and Fig. S11).

These pathways are in phase with those relating to lipid handling and insulin-related metabolic processes

in IVDAs and BAT (Fig. 3D). Adipocyte ECM composition and ECM-cell signaling directly regulate

metabolic functions such as insulin-stimulated glucose uptake, adipogenic capacity, and lipid handling,

and ECM reprograms adipocyte metabolic phenotypes in a depot-specific manner (Strieder-Barboza et al.,

2020; Baker et al., 2017). While circadian regulation of ECM-related pathways has been described in

other tissues as a dynamic interface between clocks and cellular function (Dudek et al., 2023), its cell-

autonomous circadian regulation has not previously been reported in adipocytes.

Molecular transport is cell-autonomously circadianly regulated in IVDAs and BAT.

Molecular transport pathways were enriched for circadian genes in IVDAs (FDR ≤ 0.05). These include

vesicle-mediated transport, membrane trafficking, clathrin-mediated endocytosis, and Golgi-to-ER

retrograde transport (Fig. 4A). Intra-Golgi and retrograde trafficking pathways show the strongest

enrichment and highest circadian gene content (~45%), indicating that intracellular transport pathways are

strongly circadianly regulated in IVDAs. Consistent with the enrichment results, PSEA analysis also

shows phase-clustering of processes related to molecular transport. Large transport pathways (>100

CCGs, Fig. 4B, upper), such as membrane trafficking and vesicle-mediated transport, display more

diffuse expression of their CCGs, consistent with continuous transport demand across the circadian day.

In contrast, smaller pathways (≤100 CCGs, Fig. 4B, lower) are sharply phase-clustered, peaking primarily

during the middle to late subjective day (CT7-14), with plasma lipoprotein assembly and clearance

peaking later into the subjective night. Molecular transport pathways are phase-clustered in IVDAs and

BAT but not eWAT (Fig. 2B). Although transport pathways peak within a similar circadian window, they

are systematically phase-shifted, with BAT peaking near dawn and IVDAs peaking between CT6-11,

barring plasma lipoprotein assembly and clearance. Together, these findings indicate that molecular

transport follows a cell-intrinsic circadian program whose phase is likely shaped by systemic cues in BAT.

Transcriptional and post-transcriptional pathways are strongly phase-clustered and peak late-day

in IVDAs.

Pathway enrichment analysis also revealed that pathways related to transcriptional regulation, chromatin

organization, and RNA metabolic processes are strongly enriched for CCGs in IVDAs (Fig. 5A). These

pathways span the multiple stages of gene expression, including chromatin-modification, RNA

polymerase II-dependent transcription, and rRNA processing and RNA metabolism, including mRNA

splicing. Consistent with these enrichment results, PSEA analysis shows phase-clustering of transcription

and post-transcriptional pathways (Fig. 5B). These pathways peak sharply within a narrow circadian

window around CT10-12, corresponding to the late light (rest) phase, and nearly antiphase to the timing

of the same pathways in BAT (Fig. 2B and Figs. S9-11). This phase offset indicates that systemic cues

selectively regulate the timing of transcriptional and post-transcriptional processes *in vivo*.

Sequential waves of TF-chromatin engagement in rhythmic BAT chromatin.

To predict how rhythmicity is regulated at the chromatin level, we analyzed rhythmic BAT ATAC-seq

peaks (Hepler et al., 2022) grouped into 4-hour Zeitgeber time (ZT) bins. We then performed motif

enrichment analysis ($p \leq 1\times10^{-3}$; BH q ≤ 0.01) on these bins using a curated set of circadian transcription

factor (TF) motifs (39 TFs; 42 position weight matrices) (JASPAR, Ovek Baydar et al., 2025) that are

phase-aligned between IVDAs and at least one *in vivo* adipose depot (BAT or eWAT) which allowed us to

focus on chromatin patterns that reflect circadian regulation that is likely maintained *in vitro*. This

analysis revealed distinct temporal clusters of TF-chromatin engagement across the circadian cycle (Fig.

6A). At the global level, our analysis shows that BAT chromatin accessibility oscillates across circadian

time, with two prominent peaks in total ATAC-seq signal at ZT9 and ZT17 despite a largely stable

number of accessible regions (Fig. S13). These results suggest that it is rhythmic chromatin accessibility,

rather than widespread opening or closing of regulatory elements, that provides a chromatin framework

for phase-specific TF engagement.

Motif enrichment in the ZT5-ZT9 bins (early) is dominated by ARNT-family and bHLH-PAS motifs

(ARNT, AHR::ARNT, ARNTL), NFIL3/CREB3, and KLF family motifs. The enriched CACGTG E-box

motif is annotated as an ARNT-family motif in JASPAR but is also shared by the BMAL1::CLOCK

complex. Our analysis shows approximately 70% of the early ZT5 target genes are predicted to be bound

by BMAL1 (BMAL1 ChIP-seq peaks in iWAT, Hepler et al., 2022). It is reasonable to assume that this

early ARNT enrichment likely reflects clock-dependent chromatin engagement rather than AHR-specific

binding, consistent with clock-associated promoter priming near dawn (Menet et al., 2014).

Motifs enriched during the mid-day bin (ZT13) are dominated by nuclear receptor and transcriptional co-

regulator motifs, including PPARG::RXRA, NR2F2, TFDP1, and SP1. This enrichment occurs

temporally just before the late-day, phase clustered lipid-centric metabolic program identified within the

CT17-4 window (Fig. 3D). Motifs enriched during the late-day bins (ZT17-ZT21) are dominated by

motifs associated with stress adaptation and feedback regulation, including NFKB2, CLOCK,

MAFG::NFE2L1, and CEBPA. This enrichment coincides with the late-day metabolic program identified

within the CT17-4 peak. This enrichment profile differs from that reported by Hepler et al. because our

analysis was restricted to those TFs that are circadian and in phase between IVDAs and at least one *in*

vivo adipose dataset.

Consistent with this temporal organization, enrichment of motifs associated with transcription factors

involved in metabolic regulation aligns with the peak expression of several rate-limiting enzymes and key

metabolic control points within pathways phase-clustered in the CT17–4 metabolic peak (Fig. 3D). These

include enzymes regulating lipid handling, mitochondrial metabolism, energy metabolism, and redox

balance: *Lipe, Hadhb*, and *Acox1* (ZT5); *Idh2* and *Nampt* (ZT13); *Pfkp* (ZT17); and *Acadl* (ZT21). These

enzymes function within discrete metabolic pathways rather than acting redundantly across multiple

pathways, indicating focused circadian regulation at discrete metabolic bottlenecks and coordinated

regulation across entire metabolic networks.

To connect transcription factor (TF)-chromatin engagement with biological function, we performed

Reactome pathway enrichment analysis of genes associated with rhythmic BAT chromatin accessibility

across the two predominant accessibility peaks (ZT5/9 and ZT13-21) (Fig. S14). Early peaks are

dominated by pathways related to transcriptional regulation, RNA metabolism, and signaling. In contrast,

the late peaks are dominated by pathways relating to lipid metabolism, phospholipid biosynthesis,

mitochondrial function, ECM organization, and nuclear receptor signaling (Fig. 6B).

Spatial analysis of the Distance-to-TSS of enriched motifs across the circadian day revealed promoter-

proximal motifs dominate early (ZT5-ZT9), whereas distal enhancer-associated motifs and higher motif-

to-gene ratios increase later (ZT13-ZT17), before finally declining at the end of the circadian day (ZT21)

(Fig. S15). These results begin to construct a model where a cell-autonomous relay of TF-chromatin

interactions coordinates the temporal progression from early transcriptional and regulatory programs to

later metabolic and structural pathway engagement across the circadian cycle.

DISCUSSION

To fully understand the role(s) of adipose tissue in mammalian circadian physiology, it is necessary to pursue questions in whole animals, where the full breadth of these systemic actions and interactions can be seen (e.g. Hepler and Bass, 2023; Ribas-Latre et al., 2021; Zhang et al., 2014; Pendergrast et al., 2023). Such studies by design integrate systemic cues such as light/dark, feeding patterns, hormonal regulation, and nutritional status, which influence tissue-specific molecular clocks. However, studying adipose physiology in the context of a whole animal makes it hard to distinguish cell-autonomous regulation from systemic cues, including feedback from one to the other, and thereby potentially difficult to interpret the origins of systemic metabolic mis-regulation. To begin to address this, we generated, to our knowledge, the longest, deepest adipocyte circadian time-course reported to date. This, in combination with ECHO and MetaCycle, allowed us to identify more transcripts and clock-regulated transcripts than previously possible in adipocytes, as well as more rhythmic genes in BAT and eWAT. Using these rhythmic genes, PSEA analyses allowed insights into the nature of cell-autonomous circadian regulation of adipocyte transcriptional output. Beyond this, while it is plain to see that cell-autonomous rhythmicity in IVDAs and *in vivo* is quite pervasive, an attempt to dissect all of this output for all processes risked saying very little about very much. To avoid this and to gain insight into regulatory processes with the goal of revealing cell-autonomous vs. systemic regulation, we concentrated on pathways governing transcriptional regulation, RNA metabolism, molecular transport, ECM, and energy metabolism.

We identified far more temporally regulated processes in BAT vs IVDAs or eWAT. This increase may suggest that the clock is involved in the regulation of more processes in metabolically active tissues so as to offer a higher degree of metabolic flexibility. There is also a daily rhythm in feeding and body temperature to drive metabolic processes *in vivo*. An organism may benefit from increased circadian regulation in metabolically active tissues to offer a higher degree of metabolic flexibility due in part to fluctuations in the availability of nutrients.

Agreeing with this observation, many phase-clustered pathways detected by PSEA were shared between

IVDAs and BAT, whereas overlap with eWAT was less abundant, which suggests depot-specific

differences in the timing of their regulation (Fig. 2C). The circadian clock empowers varying degrees of

metabolic flexibility across tissues (Thurley et al., 2017). This is the other side of the same molecular

coin; a highly circadianly regulated tissue, like BAT compared to eWAT may be more susceptible to

circadian disruption, given that misalignment by systemic cues is proportional to its metabolic flexibility.

Together, the BAT chromatin accessibility data support the transcriptional similarities observed between

IVDAs and BAT and provide a regulatory framework for these shared rhythms. Rhythmic BAT

chromatin accessibility follows temporally ordered patterns that align with phase-clustered transcriptional

and metabolic programs also present in IVDAs. The correspondence between TF-chromatin engagement,

pathway enrichment, and the timing of rate-limiting metabolic enzymes suggests that core aspects of

oxidative metabolism, lipid handling, and mitochondrial function are governed by cell-autonomous

chromatin programs that are preserved in BAT *in vivo*. These findings indicate that IVDAs capture key

regulatory features of BAT metabolism, supporting the idea that shared intrinsic circadian mechanisms

underlie their similar transcriptional and metabolic organization, while systemic cues primarily modulate

timing rather than defining these programs. Differences between our motif enrichment analysis and prior

results reported by Hepler et al. reflect the use of different motif enrichment algorithms and our input list

of motifs being restricted to those for TFs that are circadian and in phase between IVDAs and at least one

in the *in vivo* dataset. Despite these differences, both analyses converge on a temporally ordered

chromatin landscape linking TF engagement with metabolic and mitochondrial pathways in BAT.

MATERIALS AND METHODS

Animals and cell culture. The inguinal stromal vascular fraction (SVF) was isolated from 8- to 12-week-

old homozygous Per2::Luc C57BL/6J mice (N=4) housed at $21\pm2°C$ under a 12:12h light/dark cycle and

free access to normal chow (PicoLab Verified - 75 IF; 23.3% PRO, 12.9% FAT, 63.7% CHO). Isolated

preadipocytes were cultured in "differentiation media" before transfer to "maintenance media". Cells were

synchronized for 2-2.5 hours with 50% "base media", 50% FBS (Gibco, lot# 2102306), and 10μM

forskolin, then switched to "maintenance media" without dexamethasone and containing 0.01% d-

luciferin (100 nM, GoldBio #LUCK-100). Rhythms in luciferase activity were monitored with a

Lumicycle to confirm clocks were synchronous across samples. Quantitative PCR of *Per2* was used to

confirm synchronicity. See Supplemental Methods and Fig. S1 for further details. Sampling for two

replicate time series began at 12 hours post-synchronization (HPS) to avoid artifacts immediately

following serum shock (Balsalobre et al., 1998) and continued every 2h with the final sample collected at

72 HPS.

RNA extraction, RNA-sequencing and quantification. Total RNA was extracted from cells using the

Qiagen RNeasy Lipid Tissue Mini Kit (Cat #74804, Qiagen) according to manufacturer's protocol.

Directional RNA-seq libraries were prepared with ≥ 2μg RNA using rRNA depletion and sequenced to a

depth of at least 60M PE reads/sample to allow transcript isoform-level analyses, following the guidelines

for detecting low-abundance and isoform-specific circadian transcripts (Li et al., 2015). Reads mapping to

any residual ribosomal rRNA were removed using SortMeRNA (Kopylova et al., 2012) and remaining

reads were mapped to the *Mus musculus* GRCm38 reference genome and quantified using RSEM version

1.3.1 to produce expression in transcripts per million (TPM). See Supplemental Methods for further

details.

Filtering of transcripts. To reduce noise and exclude unexpressed and non-productive transcripts from

our analysis, we filtered transcripts to an average Count Per Million (CPM avg) ≥ 10 across all values

from replicates and limited our analysis to transcripts that map to protein-coding genes (Fig. S2).

Noncoding transcripts were not analyzed and will be the focus of a future manuscript. This filtering

process yielded 29,020 transcripts mapping to 13,601 unique genes. For accurate comparison, we

similarly filtered BAT and epididymal WAT (eWAT) microarray data from Zhang et al., 2014 to remove

transcriptional noise and included only expressed genes. This resulted in a detected protein-coding

transcriptome of 11,249 genes for BAT and 11,624 for eWAT. For both IVDAs and *in vivo* datasets, genes

under these cutoffs were considered unexpressed and excluded from analysis.

Preprocessing and rhythmicity calling. The data for IVDAs were preprocessed with LIMBR (Crowell

et al., 2019) to remove batch effects. After filtering out noise and unexpressed transcripts (see Fig. S2 and

Supplemental Methods), screening for rhythmicity was performed using Extended Circadian Harmonic

Oscillator (ECHO; De los Santos et al., 2020). Candidate circadian genes were classified as rhythmic

using ECHO based on the following criteria: one or more of their transcripts oscillating with a period of

22-30h; oscillation type of harmonic, damped or forced; Benjamini-Hochberg -adjusted p-value ≤ 0.005.

This stringent cutoff ensures only the most strongly circadian transcripts were included in our analysis. If

more than one transcript passed the requirements for CCG calling the transcript with the lowest BH-adj p-

value was selected as the gene's representative transcript. We also used MetaCycle's "meta2d" (Wu et al.,

2016) as a complementary algorithm for rhythmicity detection, using the same rhythmicity cutoffs as

described above.

Use of PANTHER and PSEA. Methods and parameters applied for use of PANTHER (gene set

enrichment analysis) to identify pathways positively enriched for circadian genes and PSEA (to

detect phase-clustered pathways) are fully described in Methods and Supplemental File 4 and 5.

Acknowledgements

We would like to thank Shannon Soucy, Tim Sullivan and Owen Wilkins at the Dartmouth

Analytics Core for their consultation on data analysis and interpretation. We would also like to

thank Joseph Takahashi who gifted the Per2::Luc mouse prior to the SNP-assisted Speed

Congenic crossing at Dartmouth. We would also like to thank Bruce Spiegelman at Harvard

Medical School for his valuable advice and guidance in this project.

Funding sources

This work was supported by grants from the National Institutes of Health to J.C.D.

(R35GM118021), J.J.L. (R35GM118022) and A. M. F (F31DK131890), COBRE (P20-

GM113132), COBRE (P20GM130454).

Author Contributions

Conceived and designed the experiments: JMW, AMF, JJL, JCD. Performed the experiments: JMW and

AMF. Analyzed the data: JMW, AMF, PAD. Wrote the paper: JMW, AMF, PAD and JCD.

Data availability

All raw RNA-sequencing data generated in this study have been submitted to the NCBI Gene Expression

Omnibus (GEO) repository under accession number GSE202686. eWAT (epididymal white adipose

tissue) and BAT (brown adipose tissue) circadian transcriptome data generated by Zhang et al., 2014

using microarrays can be found under GEO accession GSE54652. BMAL1 ChIP-seq data

generated by Hepler et al., 2022 can be found in the GEO repository (GSE181443).

Supplemental Files

Supplemental_Methods.pdf

Supplemental_Figures.pdf

Supplemental_File_S1.xlsx

Supplemental_File_S2.xlsx

Supplemental_File_S3.xlsx

Supplemental_File_S4.xlsx

Supplemental_File_S5.xlsx

Supplemental_Table_S1.docx

Supplemental_Table_S2.docx

Figure 1. Comparison of the *in vitro* circadian transcriptome with *in vivo* data from BAT and eWAT shows widespread cell-intrinsic circadian regulation.

Figure 1. Comparison of the *in vitro* circadian transcriptome with *in vivo* data from BAT and eWAT shows widespread cell-intrinsic circadian regulation. (A) Diagram of the analysis of adipocyte circadian regulation. Primary inguinal iWAT adipocytes were harvested from Per2::Luc mice and then differentiated and synchronized *in vitro* via serum shock. Two independent time courses were generated. Replicate one and two were sampled every 2 h over 60 h starting at 12 h post-serum shock. ECHO was used to model oscillations and identify circadianly regulated transcripts. Diagram created in BioRender. Worthen, J. (2026) https://BioRender.com/vf15fbu **(B)** Heatmaps showing the relative expressions of circadian transcripts which oscillated with a period of 22-30h and a BH adj P-value of <0.005 in IVDAs, BAT and eWAT. **(C)** Phase distribution of circadian transcripts in IVDAs and brown adipose tissue (BAT) and epididymal white adipose tissue (eWAT) (*in vivo* microarray data, Zhang et al., 2014) shown in circadian time (CT). **(D)** Oscillation of clock genes in Rep1 (pink) and Rep2 (blue) in IVDAs with the ECHO fitted curve shown in black. X-axis is measured in Z-score and y-axis is shown as hours post-synchronization. IVDA = *In Vitro* Differentiated Adipocytes. PC = Protein-Coding.

Figure 2. Functional analysis reveals processes that are cell-autonomously circadian yet temporally coordinated by systemic cues. (A) Top 15 Reactome top-level groups most significantly enriched for circadian genes by PANTHER (FDR < 0.05) and significantly phase-clustered by PSEA in IVDA. **(B)** Circadian time (CT) of phase-clustered pathways grouped by Reactome top-level process (colored by sample). Point size = -log$_{10}$ (q value). Gray shading delineates the four regulatory groups described in text. **(C)** Overlap of rhythmic pathways among IVDA, BAT, and eWAT identified by PSEA. Numbers indicate shared pathway counts; histograms show distributions of phase differences (ΔCT) between IVDA and each *in vivo* depot. The IVDA-BAT histogram shows an average 8-9 h phase delay, indicating substantial temporal re-phasing *in vivo*. **(D)** Representative cell-autonomous rhythmic pathways for IVDA vs BAT (left) and eWAT (right) illustrating phase alignment (<4h) within key top-level processes. ECM pathways, although not enriched for circadian genes, were highly phase-clustered in PSEA, suggesting rhythmic coordination of a smaller, coherent gene subset**.**

Figure 3. Metabolic sub-pathways display two major circadian waves corresponding to daytime anabolism and nighttime restoration. (A) Reactome energy metabolism pathways ranked by circadian enrichment relative to background. The x-axis indicates -log$_{10}$(FDR) and dot color represents the percentage of circadian genes in each pathway. Dot size corresponds to fold enrichment. **(B)**. PSEA of Metabolism. Pathways are grouped as Large (> 100 CCGs) or Other (≤ 100 CCGs) and ranked by magnitude. Colored tiles denote mean pathway phase (CT). **(C)** Histogram of all Metabolism circadian transcripts showing two phase peaks (CT 5-16 and CT 17-4). (D) PSEA of Metabolism sub-pathways run separately for Peak 1 and Peak 2 genes. Dot plots show the pathways that were exclusive to either peak 1 (top) or peak 2 (bottom). Color scale = CT; size = magnitude.

Figure 4. Circadian regulation of molecular transport pathways. (A) Enrichment analysis of Reactome transport-related pathways showing fold enrichment versus FDR significance. The x-axis indicates -$\log_{10}$(FDR) and dot color represents the percentage of circadian genes in each pathway. Dot size corresponds to fold enrichment. Analysis done using Panther.org. **(B)** Circadian phase distribution of circadian genes (CCGs) within large (>100 CCGs) and smaller (≤100 CCGs) molecular transport pathways that were determined to be phase clustered (Kuiper q value < 0.1). Bars show the phase distribution of circadian genes color-coded by 2-hour CT bins. Numbers inside tiles indicate the PSEA calculated peak phase for each pathway. Percentages to the right indicate the fraction of total genes in that Reactome pathway that are rhythmic. Large pathways display broad phase coverage, whereas smaller pathways show sharper, phase-specific clustering.

Figure 5. Transcriptional and RNA metabolic pathways are sharply phase-clustered at CT10-12, indicating a coordinated mid-day transcriptional wave in adipocytes. (A) Dot plot showing enriched transcriptional and RNA-related pathways. The list of pathways was pruned to reduce redundancy. The x-axis indicates -log$_{10}$(FDR) and dot color represents the percentage of circadian genes in each pathway. Dot size corresponds to fold enrichment. **(B)** Bar plots showing the number and circadian phase (CT) distribution of circadian genes (CCGs) within each pathway. Pathways are divided into large pathways (>100 CCGs; upper panel) and smaller pathways (≤100 CCGs; lower panel). Bars are colored by CT, and numbers on the right indicate the percentage of total circadian genes represented by each pathway. Pathways are ordered by magnitude.

Figure 6. Sequential clusters of circadian transcription factor engagement and pathway enrichment in rhythmic BAT chromatin. (A) Heatmap showing significantly enriched circadian transcription factor (TF) motifs (BH-adjusted q < 0.01) across 4-hour Zeitgeber time (ZT) bins in BAT ATAC-seq data. Color indicates -$\log_{10}$(adjusted P-value). Motifs are grouped by the ZT bin at which enrichment is maximal, revealing discrete temporal clusters of TF engagement. **(B)** Reactome pathway enrichment analysis of genes associated with rhythmic BAT chromatin accessibility, summarized across two major temporal clusters (ZT5/9, top and ZT13/17/21, bottom). Dot plots display the top 10 enriched pathways per peak ranked by -$\log_{10}$(FDR). Dot size corresponds to fold enrichment, and color denotes % circadian genes in the pathway.

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