#### Figure S1. Zhu et al.







Figure S3. Zhu et al.



Figure S4. Zhu et al.



Figure S5. Zhu et al.



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Figure S6. Zhu et al.
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#### Figure S7. Zhu et al



#### SUPPLEMENTAL FIGURES:

# Figure S1. The eigenvalue/pencil method can robustly identify circadian genes, related to Figure 1.

(A) Venn diagram comparison of the number of circadian genes identified via the eigenvalue/pencil or the COSOPT method described in (Hughes et al., 2009). Core clock genes were identified by both methods. (B-E) The distributions of the periods of the dominant oscillations from the 592 circadian genes only identified by the COSOPT method (B), the 1,486 commonly found circadian genes (C), the 3,519 circadian genes only identified by the eigenvalue approach (D) or the dominant 1,981 circadian genes identified by the eigenvalue approach (E). (F,G) All 18,108 genes analyzed by the eigenvalue/pencil method are divided in two groups based upon the presence (F) or absence (G) of superimposed ~24h oscillations. For mRNA with embedded ~24h oscillations, those with the relative amplitude of 24h oscillation larger than 0.1 are further selected. Heatmaps of mRNA expression in wild-type (WT) or BMAL1<sup>-/-</sup> (BMAL1 KO) mice from a published RNA-Seq database at different CTs (Yang et al., 2016) are shown. Further, raw microarray data used for deconvolution, deconvoluted plots, identified parameters for different oscillations and mRNA expression of representative genes (Bmal1 and Wee1 represent genes with 24h oscillation and Eif2ak3 and Gmppb represent genes without 24h oscillation in WT mice) in wild-type (WT) or BMAL1 knock-out (KO) mice are shown on the right (n=4). The RNA-Seg data are double plotted for better visualization.

#### Figure S2. Eigenvalue/pencil approach identifies wide prevalence of 8h and 12h hepatic genes, which are robust oscillations, related to Figure 1.

(A-C) To tested the robustness of the identified oscillations, we compared the oscillations derived from different microarray probes against the same gene. A 60-gene cassette containing multiple microarray probes per gene (Hughes et al., 2009) was randomly selected and subjected to mathematical decomposition. We reasoned that if oscillations from a given gene were robust, then the identified period lengths should be similar, that is with smaller coefficient of variation (CV), among different probe datasets. Likewise,

identified oscillations of largely disparate periods (larger CV value) from different probes are likely to stem from technical noise and are less likely to represent real biological oscillations. Using this rationale, we found that 8h, 12h and 24h are the most robust oscillations as approximately 90% of these oscillations have CV values less than 0.1 (A-**C).** In contrast, oscillations with periods in between (60% with CV < 0.1) and with shorter periods (<7h) (only 45% with CV<0.1) were found to be less robust (A-C). Combined with the observation that oscillations of larger periods (8~24h) remain mostly unchanged as more oscillations are added by using higher reduced models (See Supplemental Experimental Procedures), we showed that in addition to the well-characterized circadian rhythm, oscillations with periods of 12h and 8h are robust rhythms with potential biological significance. Detailed figure legends: (A) Representative deconvolution of metabolism (*Pck1*, *Acly*, *Ppara*) genes mRNA expression by the eigenvalue/pencil method. For every gene, mRNA expressions detected by two different probe sets are analyzed by the eigenvalue/pencil method. Top row represents raw microarray data (Hughes et al., 2009); second and third rows plot revealed superimposed oscillations for different probe sets; and the fourth row illustrates the amplitudes and periods of different oscillations for different probe sets with the color matching the different oscillations depicted in the second and third row. (B) Coefficient of variation (CV) values of period is plotted against the mean period value for every oscillation identified from two or three microarray probe sets against the same gene. (C) Cumulative frequencies of period CV for oscillations with different periods calculated from (B). (D-G) Comparison of the 8h cycling genes uncovered by the eigenvalue/pencil and the COSOPT method (Hughes et al., 2009). Venn diagram comparison of the number of 8h cycling genes identified via the eigenvalue/pencil or the COSOPT method (D), the distributions of the periods of the dominant oscillations from the 39 commonly found 8h cycling genes (E), the 6,255 8h genes only identified by the eigenvalue approach (F) or the dominant 1,430 8h genes identified by the eigenvalue approach (G). (H-I) Comparison of the 12h cycling genes uncovered by the eigenvalue/pencil and the RAIN method (Thaben and Westermark, 2014). Venn diagram comparison of the number of 12h cycling genes identified via the eigenvalue/pencil or the RAIN method with two p value cut-offs (H), the distributions of the periods of the dominant oscillations from the 272 commonly found 12h cycling genes

with a p value cut-off of 0.01 (I), the 266 commonly found 12h cycling genes with a p value cut-off of 0.05 (J) or the dominant 3,114 12h genes only identified by the eigenvalue approach (K). (L-O) Comparison of the 12h cycling genes uncovered by the eigenvalue/pencil and the JTK CYCLE method (Hughes et al., 2010). Venn diagram comparison of the number of 12h cycling genes identified via the eigenvalue/pencil or the JTK CYCLE method with two p value cut-offs (L), the distributions of the periods of the dominant oscillations from the 27 commonly found 12h cycling genes with a p value cutoff of 0.01 (M), the 45 commonly found 12h cycling genes with a p value cut-off of 0.05 (N) or the dominant 3,490 12h genes only identified by the eigenvalue approach (O). (P-S) Comparison of the 12h cycling genes uncovered by the eigenvalue/pencil and the ARSER method (Yang and Su, 2010). Venn diagram comparison of the number of 12h cycling genes identified via the eigenvalue/pencil or the ARSER method with two p value cut-offs (P), the distributions of the periods of the dominant oscillations from the 272 commonly found 12h cycling genes with a p value cut-off of 0.01 (Q), the 162 commonly found 12h cycling genes with a p value cut-off of 0.05 (R) or the dominant 3,218 12h genes only identified by the eigenvalue approach (S). (T) GO analyses revealing topenriched SP PIR KEYWORDS and GOTERM BP FAT terms in the 760 dominant 12h genes.

## Figure S3. 12h oscillations are transcriptionally-regulated independently from the 24h circadian rhythms in mouse liver, related to Figure 2.

(A) Heatmap of RPKM normalized mRNA expression for genes without embedded ~12h oscillations in wild-type mice in wild-type (WT) or *BMAL1<sup>-/-</sup>* (*BMAL1* KO) mice from a published RNA-Seq database at different CTs (Yang et al., 2016). (B,C) Heatmap (B) and representative mRNA expression (C) from wild-type (WT) and *CLOCK* mutant (Clockm) mice under constant darkness conditions calculated from a published micro-array database (Miller et al., 2007) and graphed as the mean  $\pm$  SEM (n = 2). (D-F) We sought further evidence for the independence of 12h rhythmicity of gene expression from the circadian rhythm by examining global run-on sequencing (GRO-Seq) data in mouse liver across one diurnal cycle (Fang et al., 2014). GRO-Seq not only can accurately quantify nascent mRNA transcription, but also can identify actively transcribing enhancer

RNAs (eRNA), the level of which closely correlates with neighboring mRNA expression (Lam et al., 2014). We first confirmed that the 12h rhythm of mRNA expression is regulated at the transcription level (D, E). Next, we reasoned that if the 12h rhythm is transcriptionally regulated independently from the circadian rhythm as speculated, we should identify examples of spatial segregation of 12h and 24h cycling eRNAs in the enhancer regions of the same gene that has both 24h and 12h oscillations superimposed at the mRNA level. As expected, we found examples of spatial segregation of 12h and 24h cycling eRNAs in Acly, Fasn and Hspa8 genes, all of which have both 24h and 12h oscillations at the mature mRNA level (F). Detailed figure legends: RPKM normalized quantification of mouse hepatic RNA synthesis rate under a 12h L/D schedule calculated from a published Gro-Seq database (Fang et al., 2014). UCSC genome browser snapshots of H3K4me3, H3K4me1 and H3K27Ac levels (Koike et al., 2012) as well as Gro-Seq tracks from ZT1 to ZT22 at three-hour intervals are provided at the top. (D) UCSC genome browser snapshots (top) and quantification of the Gro-Seq signal of the RNA transcribed from the sense (pre-mRNA) and anti-sense (anti-sense RNA) strand of *Gfpt1*, Sec23b and *Eif2ak3* gene (bottom). Boxes indicate regions of RNA used for quantification. (E) Normalized quantification of additional 12h cycling mouse hepatic premRNA level under a 12h L/D schedule. (F) UCSC genome browser snapshots of H3K4me3, H3K4me1 and H3K27Ac levels (Koike et al., 2012) as well as Gro-Seq tracks from ZT1 to ZT22 at three-hour intervals and quantification of the Gro-Seq signal of the pre-mRNAs and eRNAs transcribed from Acly, Fasn and Hspa8 gene loci. Boxes indicate regions of RNA used for quantification.

#### Figure S4. Circadian clock-independent 12h rhythmicity of gene expression can be synchronized by the metabolic/ER stress cues, related to Figure 3 and 4.

(A) Raw microarray data and eigenvalue/pencil decomposition of 10 genes in forskolinsynchronized NIH3T3 cells (Hughes et al., 2009). (B) MEFs were treated with various doses of Tu for 2h and qPCR was performed on different genes 4h later. (C, D) MEFs were treated with Tu (25ng/ml) for 2h and qPCR was performed at different time points post-Tu shock. (E) MEFs were treated with a lower dose of Tu (15ng/ml) for 2h and qPCR was performed at different time points post-Tu shock. (F, G) MEFs were transfected with different siRNAs and treated with Tu (25ng/ml) for 2h and qPCR was performed at different time points post-Tu shock for *Bmal1* (**F**) and *Xbp1* (**G**) genes. (**H**) Immunoblot analysis of BMAL1 and XBP1s in MEFs transfected with different siRNAs before Tu or GD treatment. (**I**) MEFs were transfected with different siRNAs and treated with Dex (100nM) for 30min and qPCR was performed at different time points post-Dex shock. (**J**) MEFs were transfected with different siRNAs and qPCR was performed at different time points post-Dex shock. (**J**) MEFs were transfected with different siRNAs and glucose starved for 2h and qPCR was performed at different time points post-GD. Gray boxes indicate the duration of Tu shock or GD. (**K**) Quantifications of both raw and polynomial-detrended single cell lineage recordings. (**L**) Calculated relative amplitudes of single cell oscillations with dominant periods between 11 to 14h. Data are graphed as the mean  $\pm$  SEM (n=3).

## Figure S5. XBP1s transcriptionally regulates the mammalian 12h clock., related to Figure 5.

(A) Heatmaps of proteins exhibiting 12h oscillation in mouse liver generated from a published SILAC database (Robles et al., 2014) (top) and of corresponding mRNA expression (Hughes et al., 2009) (bottom) at different CTs. (B) GO analysis revealing topenriched SP PIR KEYWORDS in 12h and 24h cycling proteins. (C) Log<sub>2</sub> mean normalized level of proteins with a 12h period oscillation in mouse liver generated from a published SILAC database involved in different biological pathways (Robles et al., 2014). (D-F) Mice were entrained under 12L:12D schedule for 14 days before released into constant darkness. 36 hours later, mice were sacrificed at a 2h interval for a total of 48h. Representative western blot analysis from liver whole cell lysates (D) and quantification and eigenvalue/pencil decomposition (E) of key hepatic metabolic enzymes (n=3). (F) Quantification and eigenvalue/pencil decomposition of plasma insulin and glucose oscillations during the 48h period (n=3). Please note that both plasma glucose and insulin levels reveal robust circadian rhythmicity, with phases consistent with previously published results (Wang et al., 2011; Zhu et al., 2015). In addition, we observed an 1.55h phase delay between insulin and glucose oscillation, consistent with the kinetics of insulin release after an increase in blood glucose levels (Gerich, 2002). Intriguingly, neither plasma glucose nor insulin oscillations exhibited 12h rhythmicity. Further, insulin oscillation showed a more "pulsatile" pattern, consisting of superimposed oscillations with periods of ~24h, 8h and 5h of relatively similar amplitudes; while the glucose oscillation is subtler with a strong circadian rhythmicity. The "pulsatile" insulin oscillating pattern across a diurnal cycle may be a macroscopic manifestation of 5~15min periodic insulin release (Porksen et al., 2002). These data indicate that mice kept under constant darkness conditions maintain a rhythmic feeding-fasting cycle. (G) Diagram of translational control of ATF4 under normal and ER stress conditions. Translation of the mammalian activating transcription factor-4 (ATF4) is regulated by two uORFs. When the eIF2-GTP-Met-tRNA ternary complex is abundant (in the presence of low levels of eIF2α phosphorylation) under normal conditions, the ribosomes initiate at uORF1 and frequently reinitiate at uORF2. As uORF2 overlaps with the ATF4 ORF, the translation of uORF2 suppresses the translation of ATF4. During ER stress, when the level of the ternary complex is reduced, the ribosome scans through uORF2 and initiates at the ATF4 initiation codon. Coding regions are shown as red rectangles, uORFs are shown as purple rectangles, ribosomes are shown in orange circles (60S subunit, large circle; 40S subunit, small circle). The ternary complex is shown as light blue circle.  $m^7G$ , cap structure. (H) Hepatic XBP1s binding to regulatory regions of key 12h metabolic genes under constant darkness conditions assayed by ChIP-qPCR (n=3~4) overlaid with RPKM normalized quantification of mouse hepatic RNA synthesis rate under a 12h L/D schedule calculated from a published Gro-Seq database (Fang et al., 2014) and BMAL1, CLOCK, PER2, H3K4me3, H3K4me1 and H3K27Ac cistromes (Koike et al., 2012). Amplicons of ChIPqPCR and regions of RNA for quantification are illustrated by arrows of different color. (I) ChIP-qPCR of XBP1s on 4.5S RNA, Per2 and Cry2 promoter regions were used as negative controls (n=3). (J-L) ChIP-qPCR of XBP1s in 2h Tu-treated MEFs with different siRNA knock-down on Sec23b (J), Eif2ak3 (K) and Hspa5 (L) genes (n=3). Data are graphed as the mean ± SEM for glucose and insulin quantification and ChIP-qPCR.

# Figure S6. Spatio-temporal coordination of the 12h rhythm of hepatic metabolism, related to Figure 5 and 6.

(A) Key carbohydrate, nucleotide and lipid metabolic pathways having rate-limiting enzymes with 12h oscillations of expression in mouse liver are demonstrated in different colors. The phases of the oscillations of metabolic enzymes in corresponding metabolic

pathways are shown in the bottom right corner. In addition, the phases of the 12h rhythm of mRNA transcription based upon Figure1 H are also shown. Black-boxed phases indicate the phase of the larger peak that is often superimposed by the peak of the 24h circadian rhythm. Metabolic pathways lacking black-boxed phases indicate that the two peaks of the 12h rhythm are symmetrical and therefore suggest that they largely lack a superimposed circadian rhythm. Key rate-limiting enzymes with a 12h rhythm are numbered from 1 to 24 and numbers in blue indicate proteins whose corresponding mRNAs do not exhibit ~12h oscillation per the eigenvalue/pencil method. Please be noted that of the 18 NUDF family members, Ndufa11, Ndufab1, Ndufb11, Ndufb2, Ndufb4 and Ndufc2 do exhibit ~12h oscillation at the mRNA level (See Table S1). Further, the subcellular localizations (mitochondria, ER or nucleus) of different metabolic pathways are depicted by blue box, red box and yellow circle, respectively. (B) 12h rhythm of RBKS mRNA and protein expression is associated with the 12h oscillation of ribose levels in mouse liver. RBKS catalyzes the conversion from D-ribose to D-ribose 5-phosphate (top). Oscillations of *Rbks* pre-mRNA and eRNA transcription (middle left), mature *Rbks* mRNA expression and its eigenvalue/pencil deconvolution (middle right), RBKS protein level (bottom left) and ribose levels (bottom right) in mouse liver. (C) 12h rhythm of SUCLA2 mRNA and protein expression is associated with the 12h oscillation of succinyl-CoA level in mouse liver. SUCLA2 catalyzes the conversion from succinyl-CoA to succinate (top). Oscillations of Sucla2 pre-mRNA transcription (middle left), mature Sucla2 mRNA expression and its eigenvalue/pencil deconvolution (middle right), SUCLA2 protein level (bottom left) and succinyl-CoA levels (bottom right) in mouse liver. (D-H) Mice under ad *libitum* (n=7) or restricted feeding (n=8) conditions are housed in CLAMS system under a 12h L/D schedule. Averaged real-time food intake of mice under ad libitum (D) or restricted feeding (E) conditions. (F) Average daily food intake in light (ZT0~12) and dark (ZT12~ZT24) phase for mice housed under *ad libitum* and restricted feeding conditions. (G-H) Eigenvalue/pencil method deconvolution of RER data from 7 individual mice housed under ad libitum conditions (G) and from 8 individual mice housed under restricted feeding conditions (H).

## Figure S7. 12h CREMA is conserved in crustaceans and nematodes, related to Figure 7.

(A) RPKM normalized quantification of mouse hepatic mtDNA-encoded RNA synthesis rate under a 12h L/D schedule calculated from published Nascent-Seg dataset (Menet et al., 2012). UCSC genome browser snapshot of Nascent-Seq tracks across the 16.1 kb mouse mitochondrial genome. Protein-coding genes encoded on the positive (H) strand are shown in red while genes on the negative (L) strand are shown in blue. Boxes show UCSC tracks with higher magnification for specific regions. (B) mRNA levels of several mtDNA-encoded genes as calculated from a published microarray database (n=5) under a 12h L/D cycle (Eckel-Mahan et al., 2013). The data are double plotted for better visualization. (C) gPCR analysis of four mtDNA-encoded genes under constant darkness conditions (n=3~7). (D) Raw data and eigenvalue/pencil decomposition of genes involved in mitochondrial gene regulation as calculated from the 48h microarray database (Hughes et al., 2009). (E) Log<sub>2</sub> (fold change to mean) for key C. elegans ortholog of mammalian 12h cycling ER and metabolism genes under temperature entrainment and free-run conditions. Two independent replicates (corresponding to experiment 3 and 4 in the original study) were shown. (F) List of top ten 12h cycling mouse genes identified by the eigenvalue/pencil method and their C. elegans orthologs with the status of 12h rhythmicity shown. Only genes with decay rate greater than 0.9 were selected. (G) Log<sub>2</sub> (fold change to mean) for C. elegans ortholog of mammalian top ten 12h cycling genes under temperature entrainment and free-run conditions. Two independent replicates (corresponding to experiment 3 and 4 in the original study) were shown. (H) List of top ten 24h cycling mouse genes identified by the eigenvalue/pencil method and their C. elegans orthologs with the status of 24h rhythmicity shown. Only genes with decay rate larger than 0.9 were selected. (I) Log<sub>2</sub> (fold change to mean) for *C. elegans* ortholog of mammalian top ten 24h cycling genes under temperature entrainment and free-run conditions. Two independent replicates (corresponding to experiment 3 and 4 in the original study) were shown.