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applications in circadian medicine A database of tissue-specific rhythmically expressed human genes has potential

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Body timing

rhythms should be considered when determining therapeutic interventions. metabolism. The data represent a useful resource for circadian medicine and strengthen the notion that circadian rhythmically expressed genes code for known drug targets or for proteins involved in drug transport and (CYCLOPS) to create a database of cycling genes in 13 human tissues. The authors show that several been only recently appreciated. Now, Ruben et al. used an algorithm called cyclic ordering by periodic structure known for more than a decade, the relevance of the circadian clock in drug response and therapeutic outcome has Although the existence of circadian clock−dependent modulation of gene expression in humans has been

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Supplementary Materials for

A database of tissue-specific rhythmically expressed human genes has potential applications in circadian medicine

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The PDF file includes:

Fig. S1. Pipeline for sample selection and CYCLOPS ordering for 13 tissues.

Fig. S2. Numbers of cycling genes by different statistical thresholds and noncoding gene type.

Fig. S3. Phase set enrichment in human and mouse counterpart tissues.

Fig. S4. Gene expression traces for top cardio-cycling drug targets.

Legends for data files S1 to S3

Other Supplementary Material for this manuscript includes the following:

(available at www.sciencetranslationalmedicine.org/cgi/content/full/10/458/eaat8806/DC1)

Data file S1 (Microsoft Excel format). Cosinor regression output for all genes and all tissues. Data file S2 (Microsoft Excel format). Set of 54 genes that cycle in 8 or more of the 13 tissues. Data file S3 (Microsoft Excel format). Set of 136 cardio-cycler drug targets.

Fig. S1. Pipeline for sample selection and CYCLOPS ordering for 13 tissues. From the GTEx collection, we selected tissues with at least 160 samples *and* evidence of intact molecular clocks. When the mammalian circadian clock is progressing normally, core clock genes show characteristic temporal sequences in expression *[\(3, 17, 51\)](https://paperpile.com/c/56DBNV/2B20+07Fd+ozW9)*. **i.** To verify presence of hallmark core clock gene phase relationships in each tissue, we calculated pairwise correlation matrices (similar to the approach in *[\(17\)](https://paperpile.com/c/56DBNV/07Fd)*). This filtering step left us with 13 tissue types. **ii.** To further

enrich for samples with robust intact clocks at time of death, we selected a subset of 160 samples for each tissue that yielded the strongest clock gene correlation matrix following 100K random samplings. This favored sample sets that were on average younger, died suddenly, and had shorter times between death and tissue stabilization. Despite this, tissue sets retained extensive diversity (**Table 1**). **iii.** 160 samples from each tissue were CYCLOPS-ordered following parameters in Anafi et al., with deviations described in Methods. **iv.** To test for gene-level rhythms, cosinor regression was run on Transcripts per Million (TPM) values as a function of estimated sample phases for each of the top 15K expressed genes in each tissue. **v.** CYCLOPSestimated phase relationships were analyzed for known clock genes to validate the quality of sample ordering.

Fig. S2. Numbers of cycling genes by different statistical thresholds and noncoding gene type. (**A**) Cycling genes detected over a series of FDR thresholds, with no boundaries on rAmp or \mathbb{R}^2 . (**B**) Cycling genes detected over a series of peak-to-trough thresholds, requiring FDR < 0.05. **(C)** Cycler counts at FDR < 0.05, rAmp \geq 0.1 and $R^2 \geq 0.1$. **(D)** Distribution of noncoding cycling genes by class (FDR < 0.05, rAmp \geq 0.1 and $R^2 \geq 0.1$). (E) Distribution of peak phases

for cyclers (FDR < 0.05) at four different peak-to-trough ratio (ptr) cutoffs. The cutoffs include cyclers above the: 25th, 50th, 75th, or 95th percentiles for ptr, within each tissue.

Fig. S3. Phase set enrichment in human and mouse counterpart tissues. PSEA was run on rhythmically-expressed genes in human (FDR < 0.05, rAmp \geq 0.1, $R^2 \geq$ 0.1) and mouse-matched (JTK Q-value < 0.05) tissues. Peak phases of expression are plotted for the top 20 pathways that were identified as phase-enriched (Kuiper's test, p-value ≤ 0.05) in both species. Phase of *ARNTL* (human) or *Arntl* (mouse) was set to 0 for comparison. Human:mouse tissue comparisons were heart-atrium:whole heart; aorta:aorta; lung:lung; visceral fat:white fat.

7.5

 5.0

 $2F$

80

60

40

 20

 10.0

 7.5

 5.0

 25

250

200

150

100

50

20

 15

 10

160

120

80

 Δ C

200

150

100

20

 15

 10

400

300

200

100

 2π

 Ω

0

 π

500

 Ω

 \overline{O}

 2π

 π

 2π

SLC6A1

 $R = 9.43$

Artery Coronary

UGCG

Heart Atrial

 $\dot{\pi}$

 $R^2 = 0.36$
ptr= 6.82

80

60

40

 20

 Ω

60

40

 20

 $\mathbf 0$

25

 20 **A**

15

10

200

150

100

50

 10.0

 7.5

5.0

25

 $0₀$

20

15

 $10¹$

5

 Ω

50

 40

30

20

 10

 $40\,$

 30_o

20

 10

 Ω

150

100

50

 $\overline{0}$

600

400

200

150

100

50

 $\overline{0}$

 Ω

SLC2A3

Heart Atrial

UGCG

Artery Coronary

 π

 $R^2 = 0.36$
ptr= 4.09

R \circ

 2π

 $R = 8.36$

 $0 - 25$

Expression (TPM)

Fig. S4. Gene expression traces for top cardio-cycling drug targets. Top 60 cardio-cycling drug targets by amplitude. Coefficient of determination (R^2) and peak-to-trough ratio (ptr) are indicated for each gene. A more permissive FDR threshold (≤ 0.1) was applied here to detect potentially noisier, but high amplitude oscillations.

Data file S1. Cosinor regression output for all genes and all tissues. Provided as a separate excel file.

Data file S2. Set of 54 genes that cycle in 8 or more of the 13 tissues. Provided as a separate excel file.

Data file S3. Set of 136 cardio-cycler drug targets. Provided as a separate excel file.