Chapter 15

Metabolites as Clock Hands: Estimation of Internal Body Time Using Blood Metabolomics

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Abstract

The circadian clock governs body time and regulates many physiological functions including sleep and wake cycles, body temperature, feeding, and hormone secretion. Alignment of drug dosing time to body time can maximize the pharmacological effect and minimize untoward effects. Therefore, a simple and robust method for estimating body time is important for drug efficacy or "chronotherapy". We previously reported that a metabolite timetable could estimate body time with good accuracy. A metabolite timetable was constructed by profiling metabolites in human and mouse with capillary electrophoresis-mass spectrometry (CE-MS) and liquid chromatography–mass spectrometry (LC-MS). In this chapter, we describe practical methods to profile and identify oscillating metabolites.

Key words Chronotherapy, Circadian, Metabolomics, Capillary electrophoresis-mass spectrometry, Liquid chromatography–mass spectrometry

1 Introduction

Mammals, as well as other many organisms, have an internal 24 h period rhythm called a circadian rhythm, which is generated through cellular transcriptional and translational feedback loops of core-clock genes in the body $[1-3]$. Circadian rhythms control many physiological functions and metabolic processes $[4]$. For example, hormone secretion is under circadian rhythm control. The abundances of melatonin and cortisol in blood, for example, oscillate in a circadian manner to regulate the daily functions of peripheral organs $[5-10]$.

There is increasing evidence that dosing time affects drug efficacy and toxicity in humans $[11-17]$. Appropriately timed administration of two anticancer drugs in ovarian cancer patients (adriamycin at 6:00 AM and cisplatin at 6:00 PM) resulted in lower toxicity than arrhythmic administration $[18]$. However, it has been reported that individual internal body times can vary up to 5–6 h among healthy humans, and even $10-12$ h for night-shift workers $\lceil 19-22 \rceil$.

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 Fig. 1 The concept of body-time estimation based on a metabolite timetable method inspired by Linnaeus' flower clock $[23]$ reduces sampling costs. The conventional way is to monitor a single molecule over a few days (highlighted by a *blue* background). By contrast, the molecular timetable method measures multiple molecules at a few time points (highlighted by an *orange* background). Then, body-time is estimated by comparing the measured profile of multiple molecules to the timetable that was constructed in advance

The conventional method to estimate body time is measure the blood level of melatonin or cortisol over 24 h because the abundance of these hormones fluctuates in a circadian manner $[5, 8, 9]$. However, this requires periodic blood sampling under a controlled environment, which severely restricts patients' activity. A simpler method is needed to measure a patient's body time in order to optimize drug efficacy $[11-13, 15, 16]$. We previously demonstrated that a molecular metabolite timetable could solve these issues [5, 23]. Our molecular timetable was inspired by Linnaeus' flower clock. In Linnaeus' flower clock, the opening or closing pattern of different flowers indicated the time of day. Likewise, we constructed a timetable of physiological molecules (e.g., metabolites) at particular times of day, and showed that the timetable could accurately estimate body time (Fig. 1).

For constructing the timetable, we used metabolomic techniques to profile hundreds to thousands of molecules in a sample. Metabolomic profiling by mass spectrometry has been used previously in other research fields including studies of bacteria, plants, and mammals $[24-26]$. The recent integration of capillary electrophoresis and mass spectrometry (CE-MS) has achieved high resolution with short measurement times [24, 27]. CE-MS is capable of identifying several hundred metabolites in a small sample volume (on the order of nanoliter). However, CE-MS, is not suited to measure neutral metabolites such as fatty acids and sugars because the technique is optimized for ionic molecules. Instead, liquid chromatography–MS (LC-MS) fills the gap left by CE-MS $[24]$.

In this chapter, we describe how we constructed a metabolite timetable for mice and humans by using CE-MS and LC-MS and provide practical lab tips $[5, 23]$.

2 Procedures

3 Materials and Sampling

3.1 Subjects and Sampling

In the mice experiment, CBA/N and C57BL/6 were used. Young mice were 5–6 weeks old, and adult mice were about 6 months old. All mice were kept under light–dark condition (light: 12 h, dark: 12 h) for more than 2 weeks, and sampled under both light– dark and constant darkness conditions. We used blood plasma for

metabolite profiling. The trunk blood was collected in tubes containing Novoheparin (Mochida Pharm) every 4 h for 2 days (12 samples in total). The blood from 4 to 10 mice per time point was mixed to obtain enough volume for constructing the metabolite timetable. The blood individual mice were used for body time measurement. Mice were fasted by removing food pellets from their cages at the light-off point one prior day to sampling to cancel the effect of food intake. Collected blood samples were quenched on ice (*see* **Note 2**) and centrifuged to remove cell products. For LC-MS samples, acetonitrile was added and strongly shaken, then supernatants were collected and lyophilized. The dried samples were stored in the freezer (−80 °C) until use (*see* **Note 3**). For CE-MS samples, 1.8 ml of methanol containing 55 μM each methionine sulfone and 2-Morpholinoethanesulfonic acid were mixed with plasma (100 μl) as internal standard and strongly shaken. Next, 800 μl of deionized water and 2 ml of chloroform were added. They were centrifuged at $2500 \times g$ for 5 min at 4 °C. The supernatant aqueous layer (800 μ l) was filtered using 5-kDa cutoff filter (Millipore) to remove cell products and proteins. Solvents were evaporated and stored in the freezer (−80 °C) until mass spectrometry analysis.

In the human experiment, six healthy volunteers (male, 20–23 years old, average 21.5 years old) participated in this study. They had no previous history of psychiatric or severe physical disease. They stayed in controlled laboratory conditions without time cues (i.e., no clock, radio, newspaper, mobile phone, or Internet access). Meals (200 kcal) were provided every 2 h and blood was collected every 2 h, using an i.v. catheter placed in a forearm vein of the subjects (*see* **Note 4**). To separate plasma, blood was centrifuged twice at $1000 \times g$ for 5 min at 4 °C. The supernatant was stored at −80 °C until mass spectrometry analysis.

4 Reagents, Instruments, and Settings for Metabolomics

We used a mixture of about 500 chemicals as internal standards. All chemical standards were purchased from commercial sources. They were dissolved in Milli-Q water, 0.1 N NaOH and stocked as 10 or 100 mM. Stock solutions were diluted to 20 or 50 μM. All CE-MS experiments were performed by using Agilent CE capillary electrophoresis system (Agilent technologies), Agilent G3250AA LC/MSD TOF system, Agilent1100 series binary HPLC pump, G1603A Agilent CE-MS adapter, and G1607A Agilent CE-ESI-MS sprayer kit. LC-MS experiments of mice were performed by using Agilent 1100 series HPLC loaded with the ZORBAX SB-C18 RRHT (Agilent Technologies). Mass spectral data was acquired by using Qstar XL mass spectrometry (Applied Biosynthesis). LC-MS human experiments were performed by using *4.1 Metabolite Standards 4.2 Instruments for Metabolome Analysis*

Agilent 1290 infinity HPLC loaded with Acquity ultra performance liquid chromatography BEH C18 column (Waters) and 6530 Accurate-Mass Q TOF LC/MS using the dual spray ESI of G-3521A (Agilent technologies) (*see* **Notes 5** and **6**).

5 Data Analysis and Statistics

5.1 Identifi cation and Quantifi cation of Metabolomics Data Data from CE-MS was analyzed by in-house software KEIO MasterHands (available on a contract basis) (*see* **Note 7**). The data analysis was performed in the following five steps: (1) Noise-filtering,

(2) Baseline-correction, (3) Migration time alignment, (4) Peak detection, (5) Integration of peak area. Migration time was normalized by an algorithm based on dynamic-programming [33]. Metabolites were identified by matching both m/z value and migration time to standard reagents, and all peaks were manually checked. Peak areas were normalized according to internal standards spiked into samples.

5.2 Screening of Time- Indicating Metabolites

First of all, we chose candidate metabolites that were constantly detected (more than 10 time points in both light–dark condition and constant darkness condition). Next, we selected metabolites that have high Pearson's correlation with a cosine curve (*see* **Note 8**). We calculated *p*-values and False Discovery Rate

Fig. 2 The circadian oscillation patterns of selected metabolites [5]. *Yellow* indicates a high level of metabolite concentration and *blue* indicates a low level in blood plasma. Metabolites were sorted by their peak time of concentration. Blood samples were collected under both light–dark (labeled with Zeitgeber time) and constant darkness (labeled with circadian time) condition. The *bar above* the graph indicates day (*white*), subjective day (*gray*), and night or subjective night (*black*)

6 Conclusion

In this chapter, we have described how a metabolite timetable was constructed for mice and humans by using CE-MS and LC-MS. We showed that our timetable method could be used in different populations, such as gender, age, and species of mice.

Our metabolite timetable method provides a robust way for minimally invasive internal body time estimation and may contribute to personalized medicine in the near future.

7 Notes

- 1. Our mouse metabolite timetable can be applied regardless of gender and age. However, some reports show that subject backgrounds, such as gender, age, and clinical history, have the potential to affect metabolomic data $[34, 35]$. Therefore, experimental design should be carefully considered depending on the purpose.
- 2. Samples should be kept cold and analyzed as quickly as possible to prevent metabolites from being degraded or modified. We recommend bringing materials such as reagents, and equipments such as a centrifuge and pipettes needed for sampling into the animal room or near the sampling space. For example, red blood cells are easily broken and can confound measurements by releasing their contents, in particular, amino acids. Samples become stable after separating plasma from red blood cells, white blood cells, and blood platelets.
- 3. When samples are stored before analysis over a long period of time (more than 1 week), it is advisable to dry up liquid and store in a deep freezer at −80 °C or in liquid nitrogen to prevent oxidation and degradation of metabolites.

Fig. 3 Circadian change of corticosterone level in mice plasma under light–dark (*left*) and constant darkness (*right*) conditions [5]. The *bar above* the graph indicates day (*white*), subjective day (*gray*), and night or subjective night (*black*). The *gray shadow* in the graph indicates the dark condition

- 4. Food intake is a primary cause of bias in blood metabolomic experiments in humans and mice, because diet directly affects body fluids $\lceil 36 \rceil$. The easiest and the most effective way to minimize the problem is fasting during time-course sampling. If fasting raises an ethical issue, we recommend providing a defined caloric amount of food at a constant time interval. In our human experiments, we served food to subjects every 2 h during sampling, and blood samples were collected just before having food $[23]$. In this way, we could select metabolites that oscillated independently of food intake.
- 5. We recommend using not only MS but also a focused approach to measure well-known oscillating metabolites. For example, measurements of melatonin or corticosterone with radioimmunoassay or enzyme-linked immunosorbent assay are beneficial to assess sample quality (Fig. 3).
- 6. There are a variety of systems, such as CE-MS, LC-MS, gas chromatography, and nuclear magnetic resonance to measure metabolites. Among them, we recommend CE-MS or LC-MS, because they are the current primary platforms for metabolomics. CE-MS is particularly appropriate for ionic metabolites and can detect a wide range of metabolites. LC-MS is appropriate for neutral metabolites such as fatty acids and sugars. We recommend using a combined platform of CE-MS and LC-MS.
- 7. There can be multiple causes of error in MS analysis. For example, capillary, column, and sheath liquid may deteriorate, and even detector sensitivity may change over the course of many measurements. It is important to occasionally, for example after every set of analysis, measure quality control samples to keep the data consistent over many samples (12 times/day \times 2 $days \times 3$ subjects = 72 samples in the case of our metabolite timetable construction). The quality control samples usually contain some reagents such as amino acids or organic acids with

predefined concentrations. This allows for canceling errors and checking sample degradation. It is also important to keep the sample tray cold (e.g., 5 °C) during analysis.

 8. We selected only metabolites which showed a gradual change in their abundance over 24 h. This is because gradual changes fit to a cosine curve provide more information about the time of day than discrete or peaky 24 h rhythms.

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