SUPPLEMENTARY MATERIAL

1. Metabolite network analysis

The network analysis is based on the method of [1]. The following data analysis is performed separately for the fasted group and the fed group. Each set of data contains 48 metabolites measured at at a baseline, and 6 sampling times, measured in hours after hemorrhage. The baseline is denoted time $t_0 = 0$, and the other sampling times are $\{t_k\}_{k=1}^6 = \{0.75, 3, 5, 9, 17, 21\}$ as described in Table 1 of the manuscript.

- Let $x_{\ell}(k)$ be the data for Metabolite ℓ at sample time t_k .
- Pre-processing:
 - The data is normalized as follows: For all ℓ , set

$$\bar{x}_{\ell}(k) = \frac{x_{\ell}(k)}{\sqrt{\sum_{k=0}^{6} x_{\ell}^{2}(k)}}$$

- The data is shifted so that the baseline measurement is 0 as follows:

$$\tilde{x}_{\ell}(k) = \bar{x}_{\ell}(k) - \bar{x}_{\ell}(0)$$

• Regression Matrix: The data from all metabolites is placed into a matrix, with each column a different metabolite, and each row a different sampling time, starting after hemorrhage and continuing until the 5th sample.

$$A = \begin{bmatrix} \tilde{x}_1(1) & \tilde{x}_2(1) & \cdots & \tilde{x}_{48}(1) \\ \tilde{x}_1(2) & \tilde{x}_2(2) & \cdots & \tilde{x}_{48}(2) \\ \vdots & \vdots & \ddots & \vdots \\ \tilde{x}_1(5) & \tilde{x}_2(5) & \cdots & \tilde{x}_{48}(5) \end{bmatrix}$$

- For n = 1 to 48:
 - The metabolite *n* is chosen as the "controlled node". The data that is to be predicted is the change in this metabolite from one sample to the next. This is denoted by $\Delta \tilde{x}_n(k)$, where $\Delta \tilde{x}_n(k) = (\tilde{x}_n(k+1) \tilde{x}_n(k)) / (t_{k+1} t_k)$. Since there are 6 samples, we can calculate $\Delta \tilde{x}_n(k)$ for $k = 1, \dots, 5$. Set

$$y = \begin{bmatrix} \Delta \tilde{x}_n(1) \\ \Delta \tilde{x}_n(2) \\ \vdots \\ \Delta \tilde{x}_n(5) \end{bmatrix}$$

- The following re-weighted regularized optimization is used to select regressors and weights that best explain the rate of change of the metabolite concentration of the controlled node. The regularization $||Wz||_1$ promotes sparsity in the solution, so that only a few elements of z will be non-zero. Select t = .95, $\delta = .001$ and length 48 vector w as $w_i = 1$ for $i = 1, \dots, 48$. For j=1 to 3, do the following
 - * Set $W = \operatorname{diag}(w)$,

$$\min_{z,\epsilon} \quad t \|Wz\|_1 + (1-t)\epsilon \\
\text{subject to} \quad \|y - Ax\|_2 \le \epsilon$$

where for length *m* vector *x*, $||x||_2 = \sqrt{\sum_{i=1}^m x_i^2}$, and $||x||_1 = \sum_{i=1}^m |x_i|$. * set

$$w(i) = \frac{\delta}{\delta + |z(i)|}$$

- Determine if $\epsilon < \frac{1}{2}(||y_{i_1}||_2 + ||y_{i_2}||_2 + \cdots + ||y_{i_m}||_2)$. If so, this is noted as a potentially interesting node. Find the indices *i* for which the weights |z(i)| > 0.1 and denote the metabolites *i* as controlling node *n*.

References

 Zavlanos, M.M., Julius, A.A., Boyd, S.P., Pappas, G.J.: Inferring stable genetic networks from steady-state data. Automatica 47(6), 1113–1122 (2011)