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1 **News & Views**

2 **SYSTEMS BIOLOGY**

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4 **Predicting and reducing hepatic lipotoxicity in non-alcoholic fatty liver disease**

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6 Utilising a systems genetics approach, a recent study identifies genetic variants and proteins
7 associated with plasma and hepatic lipid abundance and hepatic lipotoxicity.

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24 Non-alcoholic fatty liver disease (NAFLD) is a complex phenotype that results from interactions
25 between genetic, dietary, and lifestyle factors.¹ Closely associated with obesity, NAFLD is now a
26 significant public health concern worldwide, with ~25-30% of the general population in many
27 countries having fatty liver (steatosis) and ~5% having clinically meaningful fibrosis. Although several
28 genetic variants have been identified that influence susceptibility to NAFLD, as well as liver disease
29 more generally, these do not account for all of the estimated heritability of liver fat and fibrosis.¹
30 Challenging to diagnose and treat, there are unmet clinical and public health needs for non-invasive
31 biomarkers for NAFLD, and a critical research question is how to predict who is at significantly higher
32 risk for developing hepatic lipotoxicity and advanced liver disease.

33 Capitalising on a genetic reference panel of 107 inbred mouse strains, a recent study has
34 taken a systems biology approach integrating plasma and liver lipidomics with proteomics and
35 quantitative trait loci (QTL) mapping, allowing predictions to be made of genetic variants involved in
36 hepatic lipotoxicity.² Utilising replicate (n=2-3) male mice, ~60 days old, from the Hybrid Mouse
37 Diversity Panel (HMDP) strains of mice that exhibit substantial diversity in metabolic traits relevant
38 to human disease, Parker and colleagues integrate systems genetics and correlation analyses to
39 identify genetic variants and co-regulated networks that influence hepatic proteome and lipidome
40 abundance across the HMDP (Fig 1a., excerpted from Parker et al.). This work makes several
41 significant contributions to the field through identifying novel lipid-regulatory proteins, as well as
42 novel candidate circulating biomarkers for diagnosis and determining prognosis in liver disease.

43 The paper exploits the significant inter-strain variation for both liver and plasma lipid
44 abundance, with triacylglycerols, for example, varying by 27- and 16-fold respectively. Correlation
45 analyses between 190 plasma and liver lipid species, in combination with machine learning models,
46 were used to identify plasma lipid signatures that had high predictability for the abundance of hepatic
47 lipids associated with pathology (e.g. ceramides, diacylglycerol, triacylglycerol). Notably, two of
48 these signatures of plasma lipid ratios were found to correlate to human liver triacylglycerol and
49 diacylglycerol levels in a cohort of obese humans with hepatic steatosis (n=58). Integration of QTL
50 mapping for hepatic protein (pQTLs), plasma and hepatic lipid (IQTLs), and protein:lipid correlation
51 data, identified 281 single nucleotide polymorphisms mapping to 17 genetic loci that contained a
52 pQTL, an IQTL, and a direct correlation between the mapped protein and lipid in the liver. For
53 example, the GLO1 locus was identified as having both a significant cis-pQTL and having IQTLs for
54 multiple hepatic triacylglycerol species; while the GLO1 protein (glyoxalase 1) correlated with 9
55 triacylglycerol species in liver (Fig 1b, excerpted from Parker et al.). These data corroborate those
56 of Spanos and workers,³ who had, through proteomic screening, previously identified altered hepatic
57 GLO1 expression, which they confirmed in paediatric patients with NAFLD. This independent finding
58 further implicates GLO1 disruption as a causative factor in hepatic lipotoxicity and cardiometabolic
59 disease. Targeting GLO1 remains largely unexplored in the therapeutic setting. Although initial
60 results from a pilot clinical trial of a glyoxalase 1 inducer (a combination of the dietary bioactive
61 compounds trans-resveratrol and hesperetin), in participants (n=29) with overweight and obesity,

62 suggested improved glycaemic control, insulin resistance and vascular function with GLO1
63 induction.⁴

64 Global correlations between the lipidomic and proteomic datasets, in combination with
65 pathway enrichment, identified proteins associated with the proteasome and proteolysis as strongly
66 correlated with hepatic lipid abundance. The authors follow up on PSMD9, a relatively unstudied
67 protein, which they find strongly correlated to multiple hepatic and plasma lipid species. A
68 proteosomal chaperone protein, PSMD9 (also known as p27) is released during assembly of the
69 26S proteasome.⁵ Intriguingly PSMD9 has previously been shown to be associated with lipid droplets
70 under conditions of caloric restriction and intermittent fasting⁶. While early work suggested PSMD9
71 also has transactivating activity influencing glucose-dependent insulin production and glucose
72 metabolism.⁷ Parker et al. identify a significant cis-pQTL within the PSMD9 locus that also co-
73 mapped to multiple IQTLs for plasma triacylglycerols. In impressive proof-of-concept work, antisense
74 oligonucleotides were used to silence PSMD9 in two strains of mice that vary significantly in
75 abundance of liver and plasma lipids (DBA/2J and C57BL/6J), and were fed a western diet for 28
76 days. Silencing of PSMD9 reduced plasma lipids and prevented hepatic steatosis preferentially in
77 DBA/2J mice mediated in part through reductions in fatty acid synthesis by de novo lipogenesis.

78 Strengths of this work include the integration of intermediate phenotypes (proteomics,
79 lipidomics) with high-resolution genomic association mapping of the HDMP and the examination of
80 plasma and liver lipid abundance as quantitative traits. Quantifying >4,000 proteins in more than 50
81 strains of mice along with 311 lipid species across 3 classes is not a trivial experimental
82 accomplishment, and the associated datasets of Parker et al. are a rich bioinformatics resource.
83 Exploitation of the high inter-strain differences in lipid traits observed within the HDMP was a key
84 strength; without such variation, the associated genetic variant loci would not have been identified.
85 The authors rationally target their investigation a priori to a subset of known pathogenic lipid
86 subclasses, and then again to those that were most strongly correlated (ceramides, triacylglycerols
87 and diacylglycerols). Nonetheless, it is probable there are other pathogenic lipid profiles in liver,
88 along with their associated plasma signatures, which the application of unsupervised data-centric
89 approaches could help identify. Similarly, while the authors followed up on PSMD9 and highlighted
90 the independent GLO1 data, other loci identified are also worth detailed investigation and
91 examination of translation to the human disease. For example, it is interesting to note the enzyme
92 serine hydroxymethyltransferase 1 (SHMT1), was also among loci identified having both a pQTL,
93 IQTL and correlation between protein and lipid in the liver. Previous systems work utilising a genome
94 scale metabolic network in tandem with transcriptomic data from NAFLD patients suggested serine
95 deficiency in NAFLD and proposed SHMT1 as possible target.⁸

96 A key question will be in the translation to human disease. The differential effect of PSMD9
97 knockdown observed in the DBA/2J and C57BL/6J strains, underscores caution that a single target
98 for NAFLD is likely to suffice. There are, and will be further, genetic and environmental mediators
99 of hepatic lipotoxicity, particularly in outbred humans. Only male animals were profiled here; although

100 the majority of subjects in the human lipidomic dataset used in validation were female. However, it
101 should be noted these participants were undergoing bariatric surgery (thereby making the matching
102 plasma and liver lipidomics possible), and had BMIs 40-50 that further restricts the extrapolation of
103 these results. Additionally, measuring the plasma lipids represented in the ratios identified here is
104 not routine clinically, preventing the examination of existing larger scale or population datasets
105 retrospectively. Nonetheless, Parker et al provide multiple avenues for future novel cell biology and
106 their data underscore how much we have to learn about the proteasome, its role in lipid metabolism
107 and the roles of its moonlighting chaperone and component proteins.

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111 **Fig. 1 Systems genetic analysis of hepatic lipid metabolism.** Study overview (A) and back-to-
112 back Manhattan plots showing a locus significantly associated with liver triacylglycerol abundance
113 and liver GLO1 protein (B).

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