Dual extraction of mRNA and lipids from a single biological sample

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The extraction of RNA and lipids from a large number of biological samples is time-consuming and costly with steps required for both transcriptomic and lipidomic approaches. Most protocols rely on independent extraction of nucleic acids and lipids from a single sample, thereby increasing the need for biological material and inducing variability in data analysis. We investigated whether it is possible to use a standard RNA extraction procedure to analyze not only RNA levels, but also lipids in a single liver sample. We show that the organic phase obtained when using standard reagents for RNA extraction can be used to analyze lipids, including neutral lipids and fatty acids, by gas chromatography. We applied this technique to an analysis of lipids and the associated gene expression pattern in mice with hepatic steatosis induced by pharmacological activation of nuclear receptor LXR.

Both transcriptomic¹ and lipidomic² analyses have become accessible and essential tools for investigators in the field of lipid research. However, most lipid analysis and gene expression assays involve independent, costly, and time-consuming extractions.

Given that most RNA extractions are based on a phase split between chloroform and phenol, we thought it would be worth evaluating whether the organic phase can be used for complementary lipid analysis. We tested a protocol adapted from Bligh and Dyer’s extraction³ and applied it to the remaining organic phase after the aqueous phase had been used for RNA extraction. We used liver tissue and hepatic cell lines to assess the method we developed and validated the protocol by analyzing samples from wild-type mice and transgenic mice lacking liver X receptor (LXR) α and β (NR1H3 and NR1H2, respectively)⁴, two transcription factors known to play a central role in cholesterol and fatty acid metabolism⁵. By administering T0901317, a pharmacological agonist of both LXRs⁴, we induced hepatic steatosis in wild-type mice but not LXR−/− mice⁴,⁶. This steatosis occurs through an important increase in the expression of genes involved in fatty acid synthesis⁶, also called de novo lipogenesis. In this experimental setting, we tested whether LXR-induced steatosis correlates with changes in the expression of major lipid droplet proteins from the CIDE, PLIN, and FITM families⁷. In addition, sample-specific networks⁸ were analyzed, indicating Dgat2 as an atypical LXR-sensitive gene involved in triglyceride synthesis⁹,¹⁰ and down-regulated by pharmacological activation of LXR.

Results

Comparison of RNA extraction protocols. First, we compared the two RNA extraction protocols by measuring the level of various RNAs when prepared from the same liver samples. As shown in Fig. 1A, the RNA levels were similar regardless of the method. The two protocols, one relying on a commercially available reagent (TRIzol) and the other on a well-established method¹¹, were further used to investigate whether lipids could be analyzed from the remaining organic phase.

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Comparison of lipid profiles after RNA extraction. Next, we measured abundant neutral lipids, specifically triglycerides, cholesterol, and cholesterol esters, in total liver lipid extracts using different protocols (Fig. 1B). We compared the amount of these lipids to those obtained after standard lipid extraction and found no significant difference. Thus, lipid can be extracted and quantified from the remaining organic phase after RNA extraction. Consistent with this first finding, we also found that such extraction can be used for estimating total fatty acids (Fig. 1C). Interestingly, we found that, depending on the method used, the relative abundance of C18:2n-6 may be underestimated. We concluded that both commercially available reagents and standard methods can be used for dual extraction of mRNA and lipids from a single sample. To further investigate this, we performed several experiments with the commercially available reagent.

Estimation of the organic phase volume. We estimated the volume of the organic phase using radiolabeled [1-14C] palmitic acid. We checked that, under conditions of mRNA extraction with TRIzol, the upper phase contained less than 1% of the total radioactivity (Supplementary Table 1). We could also determine the organic phase volume from the amount of radioactivity contained in 25 µL carefully extracted from the organic phase.
estimating that the organic phase was approximately 560 µL (volumes used: TRIzol 1 mL, Chloroform 200 µL, data not shown). This tip is interesting in order to avoid trying to extract all of the remaining organic phase for lipid quantification.

**Linearity of RNA and lipid assays after dual extraction from a single sample.** We then focused on the RNA extraction protocol with the widely used TRIzol reagent, determining whether the protocol can be used within a range of tissue weights and cell numbers, and whether the subsequent RNA assays and lipid quantification are linear. Again, we measured total RNA and major neutral lipids from both liver samples (Fig. 2) and cultured cells (Fig. 3). Regardless of the method used for lipid extraction (Trizol or Bligh & Dyer), the amounts of the various measured lipids were correlated with the amount of tissue (Fig. 2) or number of cells (Fig. 3).

**Validation of dual extraction results.** Results obtained after dual extraction from a single sample were compared to those obtained with the classical method of Bligh and Dyer, which was used as a ref.1,4. We analyzed the results using the statistical method of Bland-Altman (Fig. 4). With one exception for each kind of lipid assay, all of the results were included in the 95% limits of agreements (bias ± 1.96 standard deviation) for both neutral lipid analysis (Fig. 4A) and fatty acid composition analysis (Fig. 4B), validating dual extraction from a single sample to quantify lipids.

**Hepatic mRNA and lipids reflecting LXR activity.** To test this newly established method in a physiological situation, we focused on the activity of the nuclear LXRα. Pharmacological activation of LXRα and LXRβ triggers marked up-regulation of the hepatic gene expression program involved in de novo fatty acid biosynthesis48. The livers from wild-type (LXR+/−) and LXR−/− mice gavaged for 5 days with a potent pharmacological LXR agonist (T0901317) were analyzed for mRNA expression and lipid content.

As expected, we observed a significant change in the expression of genes involved in fatty acid homeostasis (Fig. 5A). The expression of 75 genes related to hepatic lipid metabolism was analyzed by hierarchical clustering. We identified several clusters. Cluster 1 is a large cluster of 26 genes with up-regulated expression in response to T0901317 through a mechanism requiring LXR. This cluster contains well-established LXR target genes, including lipogenic enzymes fatty acid synthase (fas) and stearoyl-CoA desaturase 1 (Scd1) (for relative expression see Fig. 5B). In correlation with the increase in lipogenic gene expression, the liver triglyceride level was increased in wild-type mice, but not in mice lacking LXRβ (Fig. 5C). Mono-unsaturated fatty acids (MUFA) were also increased (Fig. 5D), which is consistent with the higher expression of Scd1 (Fig. 5B). Cluster 1 also contains various other genes involved in fatty acid synthesis, such as Srebp-1c, Chrebp, Acaca, and Elovl6, or in fatty acid metabolism, such as Lpcat3 and Pnpla3. In addition, this cluser contains genes involved in cholesterol disposal, such as Abcg5, Abcg8, and Gyp7a2—well-established target genes for LXR. Finally, Pparγ2, Ftm2, and Plin3 were also co-localized in this cluster of LXR-sensitive genes.

Three other distinctive patterns of gene expression grouped into other clusters. Cluster 2 contains genes with expression profiles sensitive to T0901317 but do not require LXR. Genes in cluster 3 are up-regulated in wild-type mice compared to LXR−/− mice. In cluster 4, gene expression is up-regulated in LXR−/− mice.

**Single sample network analysis.** Because this dual extraction protocol provides information on both the lipid contents and mRNA expression from a single sample, it can be used to investigate the correlation between these biological parameters. In this context, we used data from dual extraction to identify major links between gene expression and lipid homeostasis in the single-sample network (SSN) analysis developed by Liu et al.8.

We used the mRNA expression levels from the 74 genes related to lipid metabolism and 4 sets of data from lipid analysis (i.e., cholesterol, cholesterol ester, and triglyceride contents and MUFA (saturated fatty acid (SFA) + MUFA) ratio) to identify gene expression linked with specific classes of lipids. Significant changes in correlation were found compared to the reference group; all gain or loss of correlation in a pair of biological data could be correlated or anti-correlated. In this way, we could observe major perturbations induced by T0901317 treatment or the absence of LXR (Fig. 6).

Some changes in correlation observed for TO901317-treated mice (Fig. 6A) were not observed in LXR−/− mice treated with T0901317 (Fig. 6B), which suggests a major role of LXR activation in the regulation of the correlated object. For example, a loss of correlation was observed for Gpat1 (Fig. 6C) and cholesterol (Fig. 6D). In addition, in order to identify a major effect of LXR knock-out, which could be independent of T0901317 treatment, we looked at edges that appear only on ΔSSC TO901317; LXR−/− networks (Fig. 6B). Thus, using these networks, we identified a correlation between the expression profiles of Scd2 (Fig. 6E) and Dgat2 (Fig. 6F). Among the genes involved in the biosynthesis of triglycerides, Dgat2 exhibited an atypical expression profile, as it was the only gene with significantly reduced expression in response to TO901317 (Supplementary Fig. 1).

**Discussion**

In this study, we provide evidence that a standard nucleic acid extraction protocol can be adapted to analyze lipids from the same samples (Fig. 7). This protocol is reliable for preparing lipids from both cell culture and tissue samples. This extraction protocol has several benefits compared to independent extractions. First, this single extraction saves time and reduces reagent costs (Supplementary Table 2). Most importantly, this approach reduces the amount of material required and can be useful for human biopsies, or whenever a limited amount of tissue is available.

Lipid analysis is often combined with transcriptomic analysis to investigate the etiology of metabolic diseases, such as non-alcoholic liver diseases12. Non-alcoholic fatty liver disease (NAFLD) is a major public health concern worldwide13. Lipidomics is an important tool in pre-clinical14,15 and clinical studies16–19 of this continuum of pathologies associated with abnormal storage of hepatic triglycerides12. In addition to lipid metabolic profiling, the regulation of gene expression is of major interest in the field because the control of hepatic gene expression...
by several key transcription factors plays an essential role in NAFLD. LXR is an influential transcription factor in the liver and in NAFLD. In this study, LXR activity was induced pharmacologically \textit{in vivo}. We were then able to detect a correlation between LXR-dependent changes in hepatic gene expression and lipid profiles using the dual extraction protocol.

Figure 2. Trizol can be used to extract total lipids from mouse liver samples over a wide range of weights. (A) Commercially available Trizol was used to extract mRNA and (B) cholesterol, (C) cholesterol esters, and (D) triglycerides from mouse liver samples of 1 to 100 mg. (E) Bligh & Dyer was also used for cholesterol, (F) cholesterol ester, and (G) triglyceride extraction from mouse liver samples of 1 to 100 mg. RNAs were analyzed by qPCR and lipids by gas chromatography. Data are presented as means ± s.e.m. (n = 4 samples per group).
Consistent with the well-established effect of LXRs on hepatic lipogenesis\(^4\)-\(^6\), we found that a significant increase in the expression of lipogenic genes correlated with an increased accumulation of liver triglycerides. Furthermore, in accordance with an associated increase in \textit{Scd1} expression\(^2\), we also observed an increase in the abundance of hepatic MUFAs.

Figure 3. Trizol can be used to extract total lipids from hepatocytes. (A) Commercially available Trizol was used to extract mRNA and (B) cholesterol, (C) cholesterol esters, and (D) triglycerides from 1 to 5 \(\times\) 10\(^6\) JWZ cells. (E) Bligh & Dyer was also used for cholesterol, (F) cholesterol ester, and (G) triglyceride extraction from 1 to 5 \(\times\) 10\(^6\) JWZ cells. RNAs were analyzed by qPCR and lipids by gas chromatography. Data are presented as means ± s.e.m. (n = 4 samples per group).
We also used this experimental setting to question the co-regulation of other genes related to hepatic fatty acid metabolism, with a specific interest in lipid droplet-associated proteins\(^7\),\(^2\),\(^2\). These proteins are likely to influence the etiology of NAFLD, which has been associated with pharmacological\(^4\),\(^6\),\(^2\),\(^3\), genetic\(^2\),\(^4\), and nutritional\(^2\),\(^6\) regulation of LXR-dependent lipogenesis. Interestingly, we found that PPAR\(\gamma\)\(^2\) and two lipid droplet-associated proteins, namely FITM2 and PLIN3, exhibit similar gene expression pattern as lipogenic enzymes and direct LXR targets. Importantly, other proteins from the PLIN family, FITM1, and the lipid droplet-associated CIDEs, do not share this expression pattern. This finding suggests that marked up-regulation of triglyceride accumulation induced by de novo lipogenesis upon pharmacological activation of LXR activity may influence PPAR\(\gamma\)\(^2\) expression, as well specific subsets of lipid droplet proteins, such as FITM2 and PLIN3.

Finally, we used a condition-specific network analysis\(^3\) to obtain additional benefits from the dual extraction performed on single samples in these experiments. This analysis highlighted other LXR-dependent modulations of gene expression, such as the up-regulation of Gpat1 and Scd2, which are part of a large group of lipogenic enzymes regulated by LXR. Our results further confirm that TO901317 up-regulated hepatic Scd2 expression\(^2\),\(^1\) and established that LXR is required in this process. Interestingly, Dgat2\(^8\),\(^9\), which encodes the last enzyme in the pathway leading to triglyceride synthesis, appears to be an exception in this pathway, as its pattern of expression reveals an LXR-dependent inhibition in response to the LXR agonist. This inhibition of Dgat2 may represent an important rate-limiting step regulating a feedback loop in LXR-mediated triacylglycerol synthesis.

The method developed here may be particularly relevant to allow both transcriptome and lipid analysis in high throughput analysis of metabolic diseases, such as NAFLD. The approach may also be very useful for the analysis of biopsies to explore both the transcriptome and lipidome, especially when the amount of tissue is limited, and provide novel opportunities for multi-omic data analysis through sample-specific network analysis.

**Methods**

**Animals.** LXR\(\alpha\) and LXR\(\beta\) double-knockout mice (LXR\(^{-/-}\)) and their wild-type controls were maintained on a mixed-strain background (C57BL/6:129 Sv), housed in a temperature-controlled room with a 12-h light/dark cycle, and fed ad libitum with water and Global-diet 2016S (Harlan, Gannat, France). Mice of both genotypes received TO901317 (Sigma, Saint-Quentin Fallavier, France, 30 mg/kg of body weight per day) or vehicle (0.5% carboxymethyl-cellulose, 0.5% Tween80) by daily gavage for 4 days\(^3\). These mice were kindly provided by Dr. D.J. Mangelsdorf (University of Texas Southwestern Medical Center at Dallas, Dallas, TX). All experiments were performed on 19-week-old male mice.

**Ethics approval.** All experiments were approved by the relevant Animal Care and Use Committee (CEEA-86, Ministry of Research and Higher Education, France); notification TOXCOM133, and conducted in accordance with the European directive 2010/63/UE.
Figure 5. Trizol extraction of both hepatic mRNA and lipids allows detection of changes in gene expression and liver lipids in mice treated with T0901317. (A) Hierarchical classification coupled with a heatmap of hepatic gene expression. mRNA levels (74 genes related to lipid metabolism) measured by qPCR. We identified four clusters of genes differentially regulated in response to genotype or LXR activation. (B) Fas and Scd1 mRNA quantification by qPCR. (C) Liver triglycerides and the ratio of hepatic mono-unsaturated fatty acids (MUFAs) to saturated fatty acids (SFAs) + MUFAs were analyzed by gas chromatography. Data are presented as means ± s.e.m. LXR++/+ and LXR−/− mice were treated with vehicle or T0901317 (n = 5 mice per group). *P < 0.05 for the effect of T0901317 on the same genotype.
Cell culture. JWZ murine hepatic cells, also referred to as MuSH immortalized hepatocytes, were kindly provided by Dr. J.P. Gray. The cells were cultured with L-glutamine (2 mM), penicillin (100 IU/mL), streptomycin (100 mg/mL), and FBS (10%). The culture medium (DMEM, Invitrogen Cergy Pontoise, France) was supplemented with dexamethasone (1 mM).

RNA extraction. Total RNA was extracted from liver samples or mouse hepatic cells using TRIzol reagent (Invitrogen, Cergy Pontoise, France) according to the manufacturer's protocol or the standard protocol described by Chomczynski and Sacchi.

Extraction of lipid from the organic phase. After the addition of chloroform and centrifugation, the aqueous phase was retrieved and used for further RNA extraction. A sample of the organic phase obtained after RNA extraction was used for lipid extraction. We either carefully removed all or 25 µL of the organic phase and added it to a tube containing 2.5 mL of methanol and the appropriate internal standards. The internal standards were: heptadecanoic acid for fatty acid methyl esters, stigmasterol for cholesterol, cholesteryl heptadecanoate for cholesterol esters.

Figure 6. Single-sample network (SSN)-based analyses revealed a potential link between genes. LXR regulation and lipid homeostasis in mice treated with TO901317. (A,B) SSN-based analysis of the expression of 74 mRNAs (pale blue square) and 4 types of lipids (orange square). CE: cholesterol ester, Chol: cholesterol, MUFA: mono-unsaturated fatty acid, TG: triglyceride. Blue lines represent a gain of correlation, whereas red lines represent a loss of correlation, with a width proportional to the difference in correlation. Only edges of ∆SSC with P < 0.05 are presented. (C) Gpat1 quantification assayed by qPCR. (D) Liver cholesterol was analyzed by gas chromatography. (E,F) Scd2 and Dgat2 mRNA quantification assayed by qPCR. Data are presented as means ± s.e.m. LXR+/+ and LXR-/- mice were treated with vehicle or T0901317 (n = 5 mice per group). *P < 0.05 for the effect of T0901317 on the same genotype. §P < 0.05 for the effect of LXR knock-out on the same treatment.
cholesterol esters, and glyceryl triheptadecanoate for triglycerides (all from Sigma). After vortexing, the sample was filtered on glass wool prior to the addition of chloroform (2.5 mL) and water (2 mL). Next, the samples were mixed intensely by vortexing for 10 sec and centrifuged for 2 min at 1500 × g. Finally, the organic phase was extracted, evaporated to dryness, and resuspended in ethyl acetate for neutral lipid analysis or transmethylated for fatty acid analysis.

**Extraction efficiency and estimation of the average organic phase volume.** The extraction efficiency was assessed using radiolabeled [1-14C] palmitic acid (Perkin Elmer Life Sciences, Paris, France) added prior to the extraction of mRNA from 20 liver samples weighing between 20 and 120 mg. The amount of radioactivity in both phases was then measured. The total organic phase volume could be calculated from the amount of radioactivity found in 25 µL of the organic phase compared to the input measured by liquid scintillation counting (Tri-Carb 1600 TR, Packard, Meriden, CT, USA).

**Standard lipid extraction.** Triglycerides, cholesterol, cholesterol esters, and fatty acids were quantified as described previously. Briefly, following homogenization of tissue samples in methanol/5 mM EGTA (2:1, v/v), lipids corresponding to an equivalent of 1 mg of tissue were extracted in chloroform/methanol/water (2.5:2.5:2.1, v/v/v) in the presence of appropriate internal standards according to Bligh and Dyer’s method.

**Neutral lipid analysis.** Lipids extracted from the organic phase after RNA extraction or by standard lipid extraction were resuspended in 50 µL of ethyl acetate and neutral lipids analyzed by gas-liquid chromatography on a Focus Thermo Electron system using a Zebron-1 Phenomenex fused-silica capillary column (5 m, 0.32 mm internal diameter, 0.50 mm film thickness). The oven temperature was programmed from 200 to 350 °C at a rate of 5 °C/min, with hydrogen as the carrier gas (0.5 bar). The injector and detector were at 315 and 345 °C, respectively.

**Fatty acid analysis.** Fatty acids were analyzed as fatty acid methyl esters (FAMEs) as described previously. The dried lipid extract was transmethylated with 1 mL of BF3 in methanol (1:20, v/v) for 60 min at 80 °C, evaporated to dryness, and the FAMEs extracted with hexane/water (3:1). The organic phase was evaporated to dryness and dissolved in 50 µL ethyl acetate. One microliter of FAMEs was analyzed by gas-liquid chromatography on a 5890 Hewlett-Packard system (Hewlett-Packard, Palo Alto, CA) using a Famewax fused-silica capillary column (30 m, 0.32 mm internal diameter, 0.25 mm film thickness; Restek, Belfast, UK). The oven temperature was programmed from 110 to 220 °C at a rate of 2 °C/min, with hydrogen as the carrier gas (0.5 bar). The injector and detector were at 225 and 245 °C, respectively.

**Real-time qPCR.** For real-time qPCR, RNA samples (2 µg) were reverse-transcribed using SuperScript™ II reverse transcriptase (Invitrogen). Primers for SYBR Green assays are given in Supplementary Table 3. Real-time amplification was performed on an ABI Prism 7000 SDS (Applied Biosystems). All qPCR data were normalized to the mRNA levels of TATA box binding protein. Differential gene expression was calculated by the \( \Delta\Delta C_T \) method.

**Statistical analysis.** Data obtained with the different extraction protocols (Fig. 1) were analyzed using R (www.r-project.org). Data are expressed as the mean ± s.e.m. Differential effects were analyzed by ANOVA followed by Student’s t-tests with a pooled variance estimate. A P-value ≤ 0.05 was considered significant. To compare methods, Bland-Altman analyses were performed using GraphPad Prism5 software (GraphPad Software, San Diego, CA). Briefly, we analyzed the ratio of the difference between the results of dual extraction and the reference method (Bligh and Dyer) to the average of the two methods. We determined the bias ± s.d. between dual extraction and the reference method and verified that the results were included in the 95% limits of agreement (bias ± 1.96 s.d.).
Single sample network-based analysis. Initially, SSN analysis was developed to analyze transcriptomic data from a single sample compared to a large bank of reference samples. To adapt this method to our set of data, we made some modifications. First, we used the mRNA expression levels of the 74 genes related to lipid metabolism and four sets of data from hepatic lipid analysis (i.e., cholesterol, cholesterol ester, and triglyceride contents and MUFA/SFA + MUFA ratio) from LXRα−/− and LXRα−/− mice treated with or without TO901317 (n = 5 mice per group). Every calculation step was performed with R software (www.r-project.org) and a network graph generated using Cytoscape 3.5.1 software (www.cytoscape.org). After normalization to the mean of the reference groups, we analyzed the distribution of data by performing a Shapiro test. As we excluded our data following a normal distribution, we decided to apply Spearman coefficient of correlation (SCC) instead of the Pearson correlation proposed by Liu et al.8. We calculated the SCC between each pair of biological parameters in the reference group (SCCref determined on five mice). We then calculated a new SCC as proposed by Liu et al.8 by adding one mouse to the reference group (SCCref + 1) and determined differences between the two sets of SCCs (ΔSCC = SCC + 1− SCCref). We proceeded like this for all of the mice.

SSN analysis was initially developed to compare one specific sample to a large reference group, with important inter-individual variability; however, in our experiment, our reference group was composed of only five mice, with low inter-individual variability due to their common genetic strain, and it was used to make comparisons with three groups of five mice. Therefore, for these five mice per specific group, the median of five individual ΔSCC values was chosen to investigate network perturbations. The choice of the median instead of the classical arithmetic mean was made to avoid overweighting one individual amongst the five mice in a group, thereby improving the significance of the network analysis. Therefore, we finally obtained three sets of median ΔSCC values per group, a z-score was calculated and used to determine a P-value for the null hypothesis of equality between this pair of values and the mean of all pairs in this group. Based on this P-value, we selected all ΔSSC pairs including lipid values with P < 0.05. The selection finally showed genes and lipids for which there was a significant change in correlation compared to the reference group (i.e., all gain or loss of correlation in a pair of biological parameters that could be correlated, anti-correlated, or not correlated initially).

References
1. Fahy, E. et al. A comprehensive classification system for lipids. J. Lipid Res. 46, 839–61 (2005).
2. Momin, A. A. et al. A method for visualization of “omic” datasets for sphingolipid metabolism to predict potentially interesting differences. J. Lipid Res. 52, 1073–83 (2011).
3. Bligh, E. G. & Dyer, W. J. A rapid method of total lipid extraction and purification. Can J Biochem Physiol. 37(8), 911–7 (1959).
4. Repa, J. J. et al. Regulation of mouse sterol regulatory element-binding protein-1c gene (SREBP-1c) by oxysterol receptors, LXRalpha and LXRbeta. Genes Dev. 14, 2819–30 (2000).
5. Calkin, A. C. & Tontonoz, P. Transcriptional integration of metabolism by the nuclear sterol-activated receptors LXR and FXR. Nat. Rev. Mol. Cell Biol. 14, 213–24 (2012).
6. Schultz, J. R. et al. Role of LXRs in control of lipogenesis. Genes Dev. 14, 2831–8 (2012).
7. Goh, V. J. & Silver, D. L. The lipid droplet as a potential therapeutic target in NAFLD. Semin Liver Dis. 33(4), 312–20 (2013).
8. Liu, X. et al. Personalized characterization of diseases using sample-specific networks. Nucleic Acids Res. 44(22), e164 (2016).
9. Choi, C. S. et al. Suppression of diacylglycerol acyltransferase-2 (DGAT2), but not DGAT1, with antisense oligonucleotides reverses diet-induced hepatic steatosis and insulin resistance. J Biol. Chem. 31, 22678–88 (2007).
10. Yu, X. X. et al. Antisense oligonucleotide reduction of DGAT2 expression improves hepatic steatosis and hyperlipidemia in obese mice. Hepatology 42(2), 362–71 (2005).
11. Chomczynski, P. & Sacchi, N. The single-step method of RNA isolation by acid guanidinium thiocyanate-phenol-chloroform extraction: twenty—something years on. Nat. Protoc. 1, 581–5 (2006).
12. Samuel, V. T. & Shulman, G. I. Nonalcoholic Fatty Liver Disease as a Nexus of Metabolic and Hepatic Diseases. Cell Metab. 27(1), 22–41 (2018).
13. Estes, C. et al. Modeling the epidemic of nonalcoholic fatty liver disease demonstrates an exponential increase in burden of disease. Hepatology. 67(1), 123–133 (2018).
14. Chiappini, F. et al. Haptic and serum lipid signatures specific to nonalcoholic steatohepatitis in murine models. Sci Rep. 11(6), 31587 (2016).
15. Hall, Z. et al. Lipid zonation and phospholipid remodeling in nonalcoholic fatty liver disease. Hepatology. 65(4), 1165–1180 (2017).
16. Puri, P. et al. A lipidomic analysis of nonalcoholic fatty liver disease. Hepatology. 66(4), 1081–90 (2007).
17. Puri, P. et al. The plasma lipidomic signature of nonalcoholic steatohepatitis. Hepatology. 50(6), 1827–38 (2009).
18. Anjani, K. et al. Circulating phospholipid profiling identifies portal contribution to NASH signature in obesity. J Hepatol. 62(4), 905–12 (2015).
19. Loomba, R. et al. Polymunsaturated fatty acid metabolites as novel lipidomic biomarkers for noninvasive diagnosis of nonalcoholic steatohepatitis. J Lipid Res. 56(1), 185–92 (2015).
20. Duchex, S. et al. The liver X receptor: a master regulator of the gut-liver axis and a target for non alcholic fatty liver disease. Biochem Pharmacol. 86(1), 96–105 (2013).
21. Chu, K. et al. Stearyl-Coenzyme A desaturase 1 deficiency protects against hypertriglyceridemia and increases plasma high-density lipoprotein cholesterol induced by liver X receptor activation. Mol Cell Biol. 26(18), 6786–98 (2006).
22. Sahini, N. & Borlak, J. Recent insights into the molecular pathophysiology of lipid droplet formation in hepatocytes. Prog Lipid Res. 54, 86–112 (2013).
23. Duchex, S. et al. A systems biology approach to the hepatic role of the oxysterol receptor LXR in the regulation of lipogenesis highlights a cross-talk with PPARs. Biochimie. 95, 556–67 (2013).
24. Beaven, S. W. et al. Reciprocal regulation of hepatic and adipose lipogenesis by liver X receptors in obesity and insulin resistance. Cell Metab. 18(1), 106–17 (2013).
25. Duchex, S. et al. Essential fatty acids deficiency promotes lipogenic gene expression and hepatic steatosis through the liver X receptor. J Hepatol. 58(5), 984–92 (2013).
26. Duchex, S. et al. Dietary oleic acid regulates hepatic lipogenesis through a liver X receptor-dependent signaling. Plos One. 12(7), e0181393 (2017).
27. Tien, E. S. et al. Comprehensive gene expression analysis of peroxisome proliferator-treated immortalized hepatocytes: identification of peroxisome proliferator-activated receptor alpha-dependent growth regulatory genes. Cancer Res. 63, 5767–80 (2003).
28. Zadravec, D. et al. Ablation of the very-long-chain fatty acid elongase ELOVL3 in mice leads to constrained lipid storage and resistance to diet-induced obesity. Faseb J. 24, 4366–77 (2010).
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Author Contributions
N.P., S.D. and J.B.M. designed the experiments, performed experiments, analyzed the data, and contributed to writing the paper. A.P., F.L., A.M., V.L., E.F., F.S., J.M.L., S.E.S., L.L., P.G.M. and N.L. performed experiments and analyzed the data. H.G. supervised the project, designed the experiments, performed experiments, analyzed the data and wrote the paper.

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