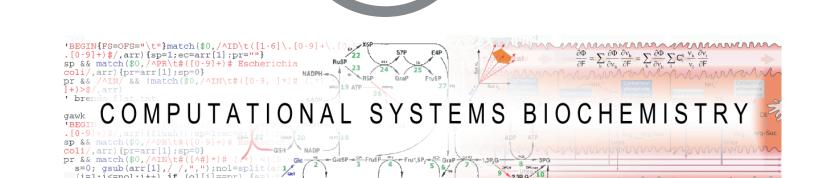
# Expression studies on cultured primary human hepatocytes with ModeScore reveal metabolic response signatures of statins and agonists of CAR, PXR, and PPARa

**Andreas Hoppe<sup>1</sup>**, Benjamin Kandel<sup>2</sup>, Mateja Hafner<sup>3,4</sup>, Damjana Rozman<sup>3</sup>, Ulrich Zanger<sup>2</sup>, Hermann-Georg Holzhütter<sup>1</sup>

<sup>1</sup>Computational Systems Biochemistry, Institute of Biochemistry, Charité, Universitätsmedizin Berlin, <sup>2</sup>Dr. Margarete Fischer-Bosch Institute of Clinical Pharmacology, Stuttgart <sup>3</sup>Institute of Biochemistry, Faculty of Medicine, University of Ljubljana, ⁴Sandoz, Biopharmaceuticals





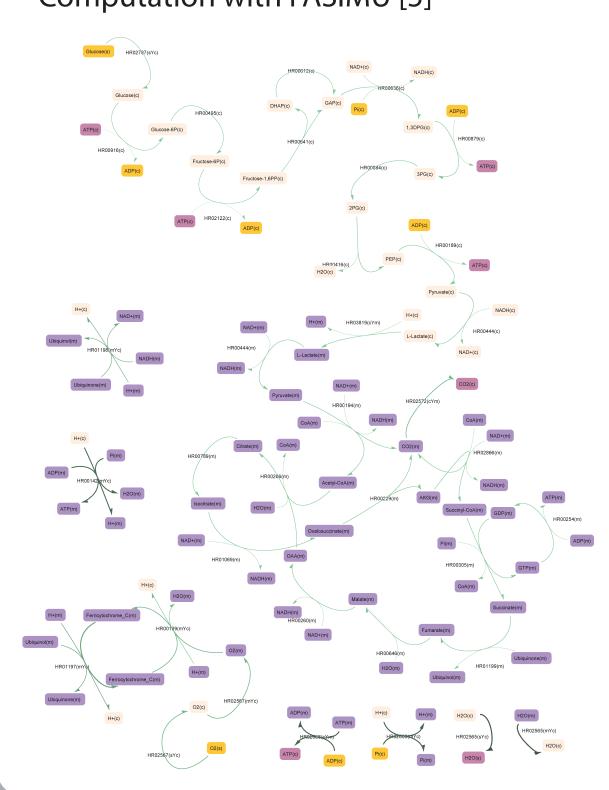
**Background.** The metabolism of hepatocytes is affected by toxins and pharmaceuticals in a multi-dimensional way. To quantify the effects with targeted metabolomics and proteomics, candidates for relevant metabolic players must be identified first, thus, the bottleneck is often the lack of initial hypotheses. Expression studies show the transcriptional response comprehensively at relatively low cost to form a basis for screening. Using the ModeScore approach it is possible to evaluate them on the level of metabolic functions.

**Results:** The novel ModeScore approach is applied to two published transcript studies of the response to Atorvastatin, Rosuvastatin, Rifampicin, and specific agonists of CAR and PXR. The method provides an enrichment for the most remarkable functions together with the genes it implements solely based on transcript profiles. Compared with the annotation-based approach, ModeScore relates the transcript changes directly to the enzymes and transporters needed for a metabolic function and not on an artificial definition, thus, present testable hypotheses on the level of cellular function. It is demonstrated how specific response patterns emerge from the ModeScore calculations and how they uncover the cellular strategy of transcriptional regulation.

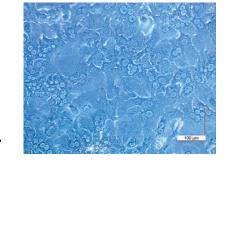
**In conclusion,** the novel method provides an enrichment for the most promising functions together with the genes it is based upon for further metabolic analysis solely based on transcript profiles.

# Reference flux distributions (modes)

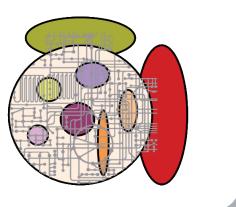
- Computation with FASIMU [5]



Data. Full genome RNA transcript profiles (Affymetrix HuGene 2.1) of primary hepatocytes from human donors cultured on collagen monolayer. Treatment with Rifampicin (PXR agonist), Atorvastatin, Rosuvastatin (24h, 48h, SterolTalk experiments [1]), Rifampicin, CITCO (CAR agonist), WY1436 (PPARα agonist) (24h, Stuttgart experiments [2]).



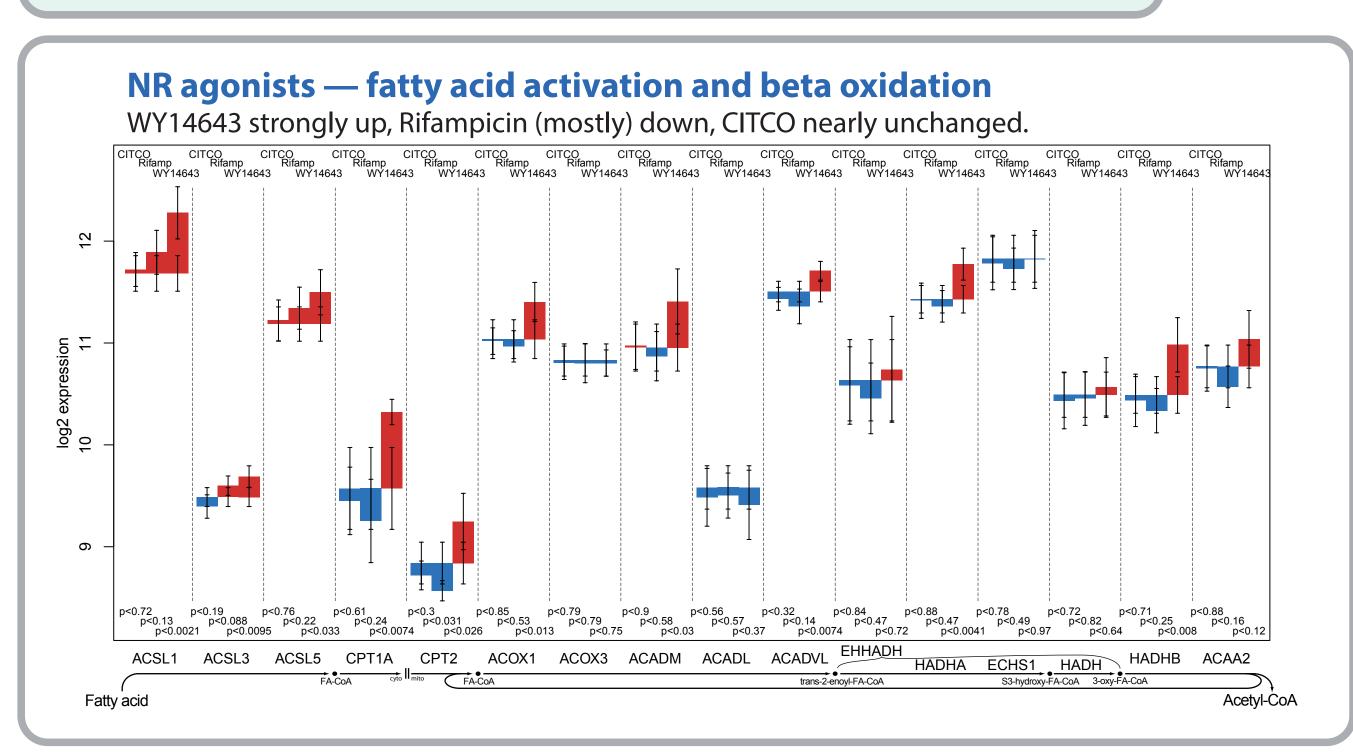
**Network.** HepatoNet1b, manually curated network of the human hepatocyte [6], refined to cover more functions, comprises 1500 localized metabolic species, 2702 reactions, 879 annotated genes [3].



#### Functions definition [3].

Plethora of metabolic functions (992), three categories:

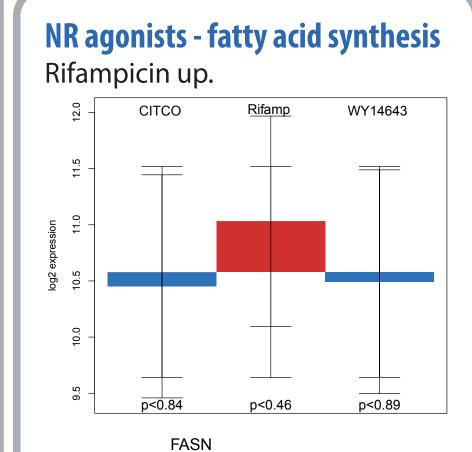
- Regeneration of important intemediates (72)
- Function of organismic duty (379)
- Synthesis and degradation of cellular constituents (541)



#### **Difference Bar charts**

treated vs. untreatedblue — down-regulation

red — up-regulationstandard deviationpaired t-test p-value



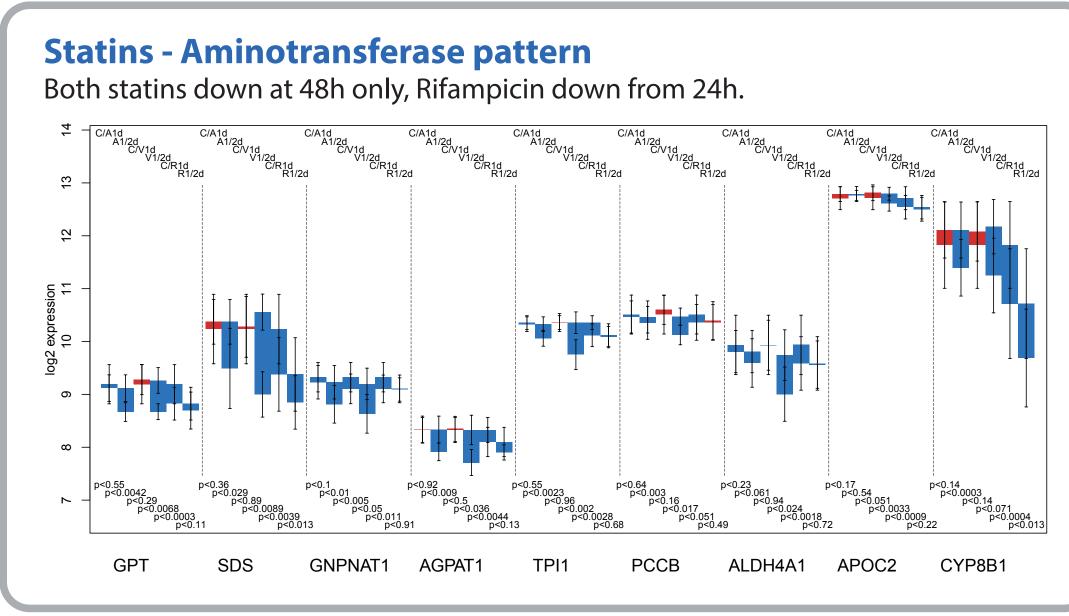
#### **Function ranks by ModeScore amplitude**

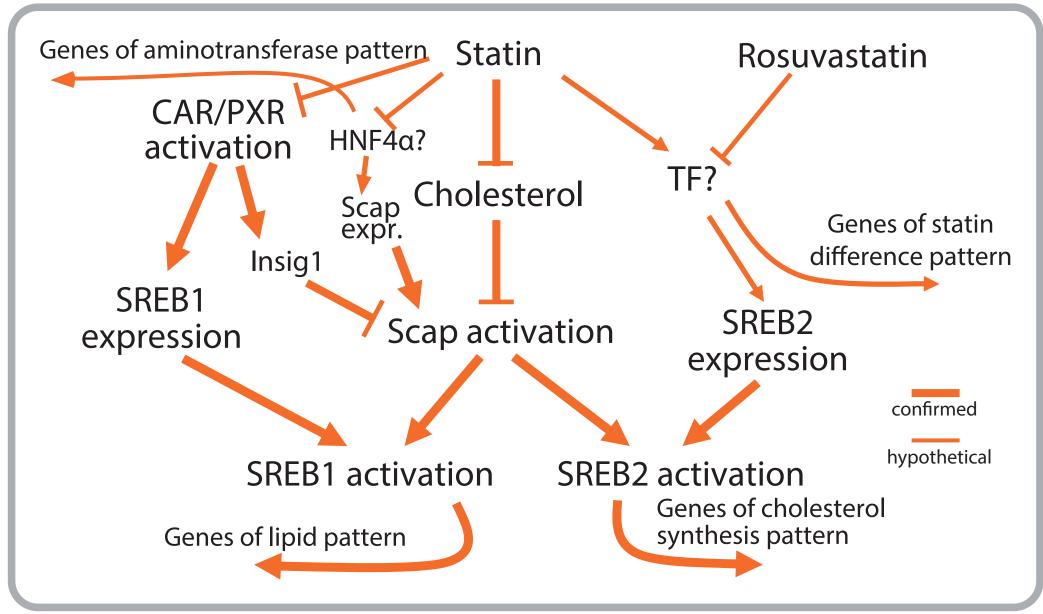
Relative expression scores treated vs. untreated. Only the most upregulated shown.

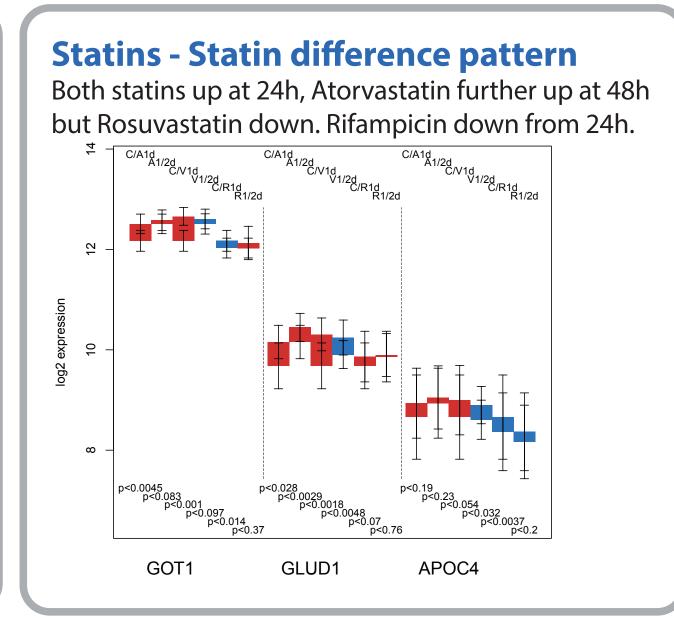
Simulation	Rifamp WY14643 ampl score		Simulation	con/Atory 4 ampl score	8h con/Rosuv 48h ampl score
109 Stearate 108 Oleate 106 Palmitate 46 Palmitolate from Arachidonate 43 Palmitolate from Palmitate 41 Taurine from Cysteine 121 PI 55 Bilirubin conjugation 125 Cholesterol 1 Aerobic ATP rephosph (FA) 9 GSH red from NADH redox pot 8 Aerobic reduction of FAD (FA) 73 Oleate degr	0.44 0.54   -0.08 0.56 0.44 0.5   -0.08 0.49 0.42 0.66   -0.08 0.62 0.41 0.53   -0.08 0.54 0.13 0.42   0.17 0.43 0.16 0.44   0.1 0.56 0.1 0.43   0.13 0.44 0.25 0.39   -0.05 0.4 0.08 0.47   0.06 0.39 -0.06 0.59   0.17 0.27	27 28 148 62 61 54 49 63 14 140 51 124 50 53	Isopentenyl-PP Farnesyl-PP Cholesterol Gly-CD-cholate(b) Glycocholate(b) Arachidonate from Dihomo- $\gamma$ -linolenate $\gamma$ -Linolenate from Linoleate Taurocholate(b) Activated methyl group (THF) CMP-N-acetylneuraminate Arachidonate from Linoleate Glycine Dihomo- $\gamma$ -linolenate from Linoleate Arachidonate from $\gamma$ -Linolenate	2.7 0.62 2.57 0.58 1.94 0.6 1.83 0.36 1.79 0.35 te 1.05 0.48 1.02 0.46 0.23 0.31 1.08 0.46 0.71 0.46 0.94 0.47 0.84 0.4 0.86 0.44 0.86 0.44	2.88 0.57 2.71 0.54 1.88 0.6 1.84 0.33 1.82 0.32 1.69 0.42 1.67 0.39 1.84 0.26 0.9 0.49 0.66 0.47 0.2 0.42 0.28 0.33 0.2 0.45 0.2 0.45

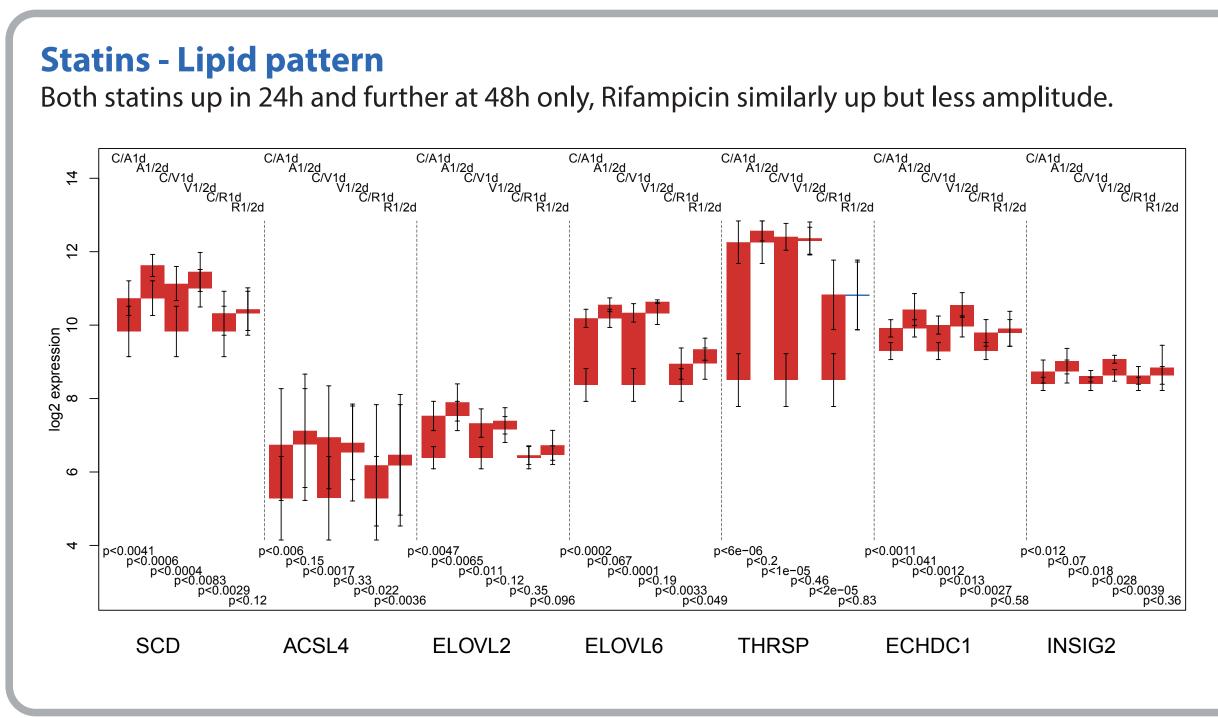
#### **Patterns**

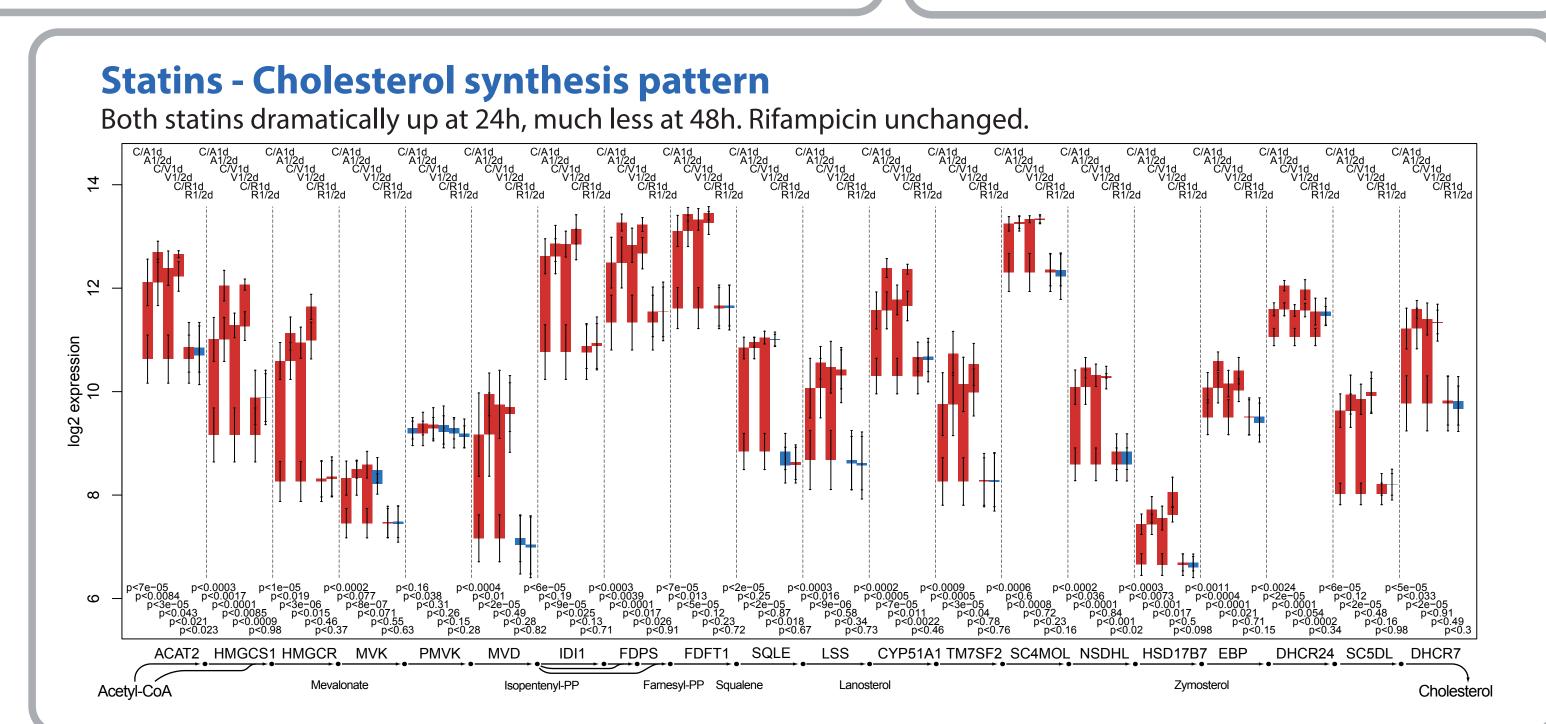
Groups of genes show a striking similarity in response to the different treatments. Some of them are obviously functionally related (Cholesterol synthesis), others less so. Presumably, they are commonly regulated. For some of them, the regulation has already been discovered (SREB1/2), for others (Statin difference) it has yet to be confirmed.













Contact:
Dr. Andreas Hoppe
Charité, University Medicine Berlin
Institute for Biochemistry
Charitépl. 1/Virchowweg 6
10117 Berlin, Germany
Phone: +49-30-450 528 176
E-mail: hoppe@bioinformatics.org

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