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Metabolic stress-induced microRNA and mRNA expression profiles of human fibroblasts

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Abstract

Metabolic and oxidative stresses induce physiological adaptation processes, disrupting a finely tuned, coordinated network of gene expression. To better understand the interplay between the mRNA and miRNA transcriptomes, we examined how two distinct metabolic stressors alter the expression profile of human dermal fibroblasts.

Primary fibroblast cultures were obtained from skin biopsies of 17 healthy subjects. Metabolic stress was evoked by growing subcultured cells in glucose deprived, galactose enriched (GAL) or lipid reduced, cholesterol deficient (RL) media, and compared to parallel-cultured fibroblasts grown in standard (STD) medium. This was followed by mRNA expression profiling and assessment of > 1,000 miRNAs levels across all three conditions. The miRNA expression levels were subsequently correlated to the mRNA expression profile.

Metabolic stress by RL and GAL both produced significant, strongly correlated mRNA/miRNA changes. At the single gene level four miRNAs (miR-129-3p, miR-146b-5p, miR-543 and miR-550a) showed significant and comparable expression changes in both experimental conditions. These miRNAs appeared to have a significant physiological effect on the transcriptome, as nearly 10% of the predicted targets reported changes at mRNA level. The two distinct metabolic stressors induced comparable changes in the miRNome profile, suggesting a common defensive response of the fibroblasts to altered homeostasis. The differentially expressed miR-129-3p, miR-146b-5p, miR-543 and miR-550a regulated multiple genes (e.g. NGEF, NOVA1, PDE5A) with region- and age-specific transcription in the human brain, suggesting that deregulation of these miRNAs might have significant consequences on CNS function. The overall findings suggest that analysis of stress-induced responses of peripheral fibroblasts, obtained from patients with psychiatric disorders is a promising avenue for future research endeavors.

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human fibroblast; stress; miRNA; mRNA; profiling; lipid; galactose; qPCR

INTRODUCTION

Postmortem brain studies of psychiatric disorders show complex transcriptome profile changes that encompass both miRNA and mRNA disturbances [1,2]. Yet, these studies face considerable confounds, limiting mechanistic and molecular investigation into the causality of the observed changes. The uniquely human occurrence of major psychiatric disorders and the absence of *bona fide* animal models [3,4,5] underscore the need for analyzing human tissue. However, the usefulness of post-mortem brain tissue is limited primarily to static measures, such as gene expression analysis. Patient fibroblasts, due to their abundance, availability, and neuron-like receptor and protein expression are appealing alternative *in vitro* models for functional analysis of brain disorders [6,7].

Stress is a strong modulator of gene expression [8]. In vitro, glucose-depleted, galactosesupplemented medium is a widely used metabolic stress treatment in fibroblast cultures [9]: glucose depletion increases the level of reactive-oxygen species (ROS), which results in elevated, compensatory glutathione production [7]. Yet, over a prolonged time starvation results in reduction of NADPH, limiting the efficacy of the glutathione protection to the cells. In contrast, reduction in the exogenous cholesterol is a different kind, but related metabolic stress: to maintain their membrane structure and survive, fibroblasts are forced to synthesize their own cholesterol, stressing their metabolic machinery [10]. These two metabolic stress models are likely to share common characteristics: there is an increased rate of glucose incorporation into lipid when cells are switched into lipid-deficient medium [11] and C^{14} galactose incorporates into galactolipids [12]. Therefore, both would be expected to manifest common disturbances in the gene expression machinery.

Micro-RNAs (miRNA) are 18-22 nucleotide long fine-tuning regulators of gene expression and protein production [13]. Thousands of miRNAs are encoded in the human genome, controlling the function of approximately 50% of our genes. miRNA expression is tissue, environment and individual specific [14]. A single miRNA can regulate the expression of hundreds of mRNAs, cause their degradation, destabilization, stocking or transport.

The present study was aimed to answer three questions. First, do the different metabolic stressors alter the miRNA expression profile of human fibroblasts? Second, are the miRNA expression profiles similar between two different metabolic stressors? Finally, do the miRNA expression changes result in changes at the mRNA level? Importantly, these studies provide critical baseline data for similar experiments that will be performed on fibroblasts from patients with schizophrenia, major depression and bipolar disorder.

MATERIALS AND METHODS

ETHICS STATEMENT

The Vanderbilt University Institutional Review Board (IRB) approved the study and written informed consent was obtained from all study participants before any procedures were conducted.

HUMAN FIBROBLAST CELL CULTURES

Specimens for fibroblast cultures were obtained by skin biopsies (~ 1×2 mm) from 17 healthy subjects (4 males and 13 females) (Supplemental Material 1) as described previously

[15]. Exclusion criteria included any DSM-IV diagnosis [16] or any medical condition that would preclude the biopsy.

All fibroblast samples used in this experiment were cultured between passages 5 and 10 using Dulbecco's modified Eagle's mediumTM (DMEM; Mediatech, Manassas, VA, USA) containing 25mM glucose, 1 mM sodium pyruvate supplemented with 2 mM L-glutamine (Mediatech), 10% fetal bovine serum (FBS; Thermo Scientific HyClone, Logan, UT, USA) and antibiotic/antimycotic solution (A/A; Invitrogen) at 37 °C and 5% CO₂. The cells were cultured simultaneously under same conditions until reaching confluence, trypsinized and counted.

A single cell line from each individual was divided into three plates (plating density 1.2×106 cells/plate) and after overnight adherence, the medium was changed into 3 different conditions. All cultures were grown for 7 days, with a media change every 2 days. (Figure 1) and were grown in: 1) lipid reduced (RL) medium (DMEM supplemented with lipid reduced FBS (Thermo Scientific HyClone, Logan, UT, USA)); 2) galactose-containing (GAL) medium (DMEM deprived of glucose (Mediatech) supplemented with 10mM galactose) and 3) standard, glucose-containing (STD) medium. The medium was changed three times a week. Cell growth and proliferation were not affected by GAL or RL treatment. After one week the cells were washed with ice-cold PBS twice, trypsinized, pelleted (700g for 8 min on 18 °C) and stored on –80 °C until the RNA isolation.

RNA ISOLATION

Total cell RNA and small RNA fractions were isolated from the frozen samples using the mirVana miRNA isolation Kit (Ambion, Austin, TX) according to manufacturer's instructions. In brief, the pellet was suspended in 600 μ l Lysis/Binding Solution, than 60 μ l Homogenate Additive was added. After 10 min incubation time for total RNA isolation 1.25× volume of 100% ethanol was added to the aqueous phase and the supernatant was purified using the proprietary solutions provided by the manufacturer. Finally the total RNA was collected with 100 μ l Elution Solution. For small RNA isolation we added on-thrid volume of 100% ethanol to the aqueous phase and two-thrid volume of 100% ethanol to the flow-through. The small RNA specied were collected from the filter with 100 μ l Elution Solution. Agilent 2100 Bioanalyzer was used to determine the quality and size distribution of the total RNA and small RNA fractions. All samples showed a RNA Intergity Number (RIN) >8.0. The samples were stored on -80 °C until used.

DNA MICROARRAY and miRNome PCR ARRAY

mRNA expression—mRNA levels were examined with the GeneChip HT HG-U133+ PM Array Plate (Affymetrix, Santa Clara, CA, USA). cDNA synthesis from total RNA, IVT, labeling and hybridization were performed according to manufacturer's protocol.

miRNA expression—From small RNA samples cDNA was prepared with miScript II RT Kit (Qiagen, Valencia, CA, USA) using miScript HiSpec Buffer. The DNA was quantified with NanoDrop ND-1000 spectophotometer, yielding 693-839 ng/µl. After reverse transcription, the individual sample cDNAs were pooled into four groups based the gender and the age. Each group contained material from 4-5 individuals. Group 1: males age of 30-48; group 2: females age of 20-25; group 3: females age of 27-40; group 4: females age of 44-52 (Supplemental Material 1). The relative amounts of 1008 miRNAs were measured with miRNome miRNA PCR Arrays using miScript SYBR Green PCR Kit (Qiagen, Valencia, CA, USA) according to the manufacturer's instructions.

VALIDATION OF mRNA FINDINGS

Based on our pathway analyses we selected a panel of 10 lipid biosynthesis genes, and validated the mRNAs using qPCR on the individual control, GAL-treated and RL-exposed samples (n=18). Total RNA (500 ng) from each sample was reverse transcribed to cDNA using RT2 First Strand Kit (Qiagen) using the instructions of the manufacturer. Custom-designed RT2 Profiler PCR Arrays (Qiagen) were provided in 96-well plate format and contained primer assays for 10 focused genes (SCD, FABP3, DHCR7, INSIG1, QPRT, HMGCR, HMGCS1, LDLR, FADS1, FADS2) and 2 housekeeping genes (ACTB and GAPDH). RT2 Profiler PCR Arrays assays were performed on the real-time cycler ABI Prism 7300 System (Applied Biosystems). The exported Cts were analyzed using SABiosciences PCR Array Data Analysis Template Excel software and compared across STD-RL and STD-GAL groups using a paired experimental design.

VALIDATION OF miRNA FINDINGS

22 miRNAs were selected for follow up on the 17 individual samples (hsa-miR-146b-5p, hsa-miR-550a, hsa-miR-129-3p, hsa-miR-214, hsa-miR-21, hsa-miR-22, hsa-miR-132, hsa-miR-376c, hsa-miR-19a, hsa-miR-195, hsa-miR-181a, hsa-miR-486-5p, hsa-miR-377, hsa-miR-424, hsa-miR-542-3p, hsa-miR-22, hsa-miR-103a, hsa-miR-376b, hsa-miR-29b, hsa-miR-185, hsa-miR-564, hsa-miR-34a). The assays was identical to the one used on the pooled samples described above.

DATA ANALYSIS

Differential expression—For the mRNA studies, after standard image segmentation of the microarray images, RMA normalization was performed using GenePattern software [17]. At the individual level, the gene was considered to be differentially expressed if it showed an absolute average log ratio |ALR| 0.378 and p-value 0.01 between the experimental and control conditions. Pathway analyses were performed using GenePattern – Gene Set Enrichment Analysis (GSEA) [18] and its BioCarta Database.

For miRNA expression quantification comparative Ct method was used with SNORD61, SNORD68, SNORD72, SNORD95, SNORD96A and RNU6-2 as housekeeping genes. A miRNA was considered differentially expressed if it showed a $|\Delta\Delta Ct| = 0.583$ and both a groupwise and pairwise p value of p = 0.05 between the experimental and control condition.

These parameters were chosen based on previously analyzed datasets [17,18], and followed up by qPCR validation.

Hierarchical clustering and calculation of correlation—For both the mRNA and miRNA data two-way unsupervised clustering using Euclidian distance was performed using the GenePattern software package [19]. For mRNA, this was performed using the RMA - normalized log_2 gene expression values from each individual sample, while for the miRNA the Δ Ct values were used. Correlations were calculated in MS-Excel 2010 using Pearson coefficient.

Identification of miRNA targets—To identify the potentially affected mRNAs by the four miRNAs that were changed in both the GAL- and RL-exposed cultures (miR-129-3p, miR-146b-5p, miR-543 and miR-550a), we searched the miRDB online database [21] and established the list of their potential targets.

COMPARISON TO BRAIN EXPRESSION

To test the relevance of our fibroblast findings in context of human brain disorders, we examined the mRNA expression of putative miR-129-3p, miR-146b-5p, miR-543 and miR-550a targets in the developing human brain [33].

RESULTS

METABOLIC STRESS-INDUCED mRNA CHANGES

Our mRNA expression profiling revealed 2063 differentially expressed genes in the GAL group compared to the matched controls. Of these, 1117 (54.1%) were up-regulated (Supplemental Material 2A), while 946 (45.9%) showed downregulation. These regulated gene products were involved in cell cycle, apoptosis, inflammatory response and mRNA processing related pathways, and in metabolic adaptation. RL treatment had also a widespread effect on the transcriptome. 984 gene transcripts were affected by the RL treatment, with 640 mRNAs increased (65.0%) and 344 (35.0%) transcripts repressed (Supplemental Material 2B). Two-way unsupervised hierarchical clustering provided a good separation of samples based on the treatment condition (Supplemental Material 3AB). Pathway analysis with BioCarta classification revealed enrichment of 18 groups of transcripts in the GAL condition (Table 1A), while in the RL condition, the analysis revealed enrichment in 12 pathways and transcript repression in 3 pathways (Table 1B). Notably, 4 of these pathways were commonly upregulated in both the GAL and RL condition: peroxisome proliferator-activated receptor α (PPARA), carbohydrate response element binding protein 2 (CHREBP2), influence of Ras and Rho proteins on G1 to S Transition (RACCYCD) and stress induction of HSP regulation (HSP27). As there was a significant enrichment of lipid biosynthesis genes in the fibroblast as a result of exposure to RL and GAL conditions, we decided to validate a panel 10 commonly upregulated genes (SCD, FABP3, DHCR7, INSIG1, QPRT, HMGCR, HMGCS1, LDLR, FADS1, FADS2). qPCR assessment of expression validated 9 out of 10 transcript changes in both RL and GAL conditions (all except HMGCR, which showed the same directionality and magnitude of change, but did not reach statistical significance) (Supplemental Material 4) and revealed a high correlation between the microarray and qPCR results (GAL-STD comparison: r=0.92; RL-STD comparison: r=0.72) (Figure 2).

miRNA EXPRESSION IN RESPONSE TO METABOLIC STRESS

We found that in human fibroblasts 45 miRNAs were differentially expressed due to the GAL treatment (Table 2 A,B) ($|\Delta\Delta Ct| = 0.583$; p = 0.05), with approximately same number of miRNAs increased and decreased (24 miRNAs and 21 miRNAs, respectively). The RL treatment affected 34 miRNAs, with more prominent downregulation than upregulation (27 and 7, respectively). When compared to the control cultures, four miRNAs showing significant differential expression overlapped in the two metabolic stress conditions (GAL, RL) (Figure 3). However, the overall pattern of miRNA profiles between the two stressors was strong: the differentially expressed miRNA's in the GAL condition, when correlated with their expression level in the RL condition, showed a significant correlation (r=0.71, p<0.001) (Figure 4A). Similarly, the expression levels of differentially expressed miRNA's in the RL condition were correlated with the expression levels observed in the RL condition (r=0.65, p<0.005) (Figure 4B). Furthermore, the combined list of these miRNA expression changes, when subjected to a two-way unsupervised hierarchical clustering, perfectly separated out the CNT, GAL and RL conditions (Figure 5). Thus, while the most changed miRNAs were not identical between the GAL and RL, the overall pattern suggested a comparable miRNA response between the two very different metabolic stressors.

As the initial screening was performed on 4 pools of samples which could theoretically bias our data, we wanted to validate the expression of 22 miRNAs on our 17 individual samples. In this follow-up study the initial miRNA findings on the pooled samples showed the expected expression pattern on the individual experimental samples (Figure 6).

CORRELATION OF miRNA WITH mRNA EXPRESSION

To establish how metabolic stress-induced miRNAs affect the mRNA transcriptome, we focused our attention on the four significantly changed miRNAs that overlapped between the GAL- and RL-treatment: miR-129-3p, miR-146b-5p, miR-543 and miR-550a. To identify the target mRNAs of these miRNAs we established the list of their potential targets using the miRDB online data base. The database revealed 211 putative mRNA targets of miR-129-3p, 243 mRNA targets of miR146b-5p, 129 mRNA targets of miR-550a, and 921 targets for miR-543 (Supplemental Material 5). Of the combined 1504 potential targets 140 mRNAs showed significant change at gene expression level (Table 3A-C). It is noteworthy that multiple target genes showed significantly altered transcription levels in both the GAL-and RL-treated human fibroblast cultures, further arguing for a consistent effect of the two different metabolic stressors.

Finally, we attempted to establish if any of these putative mRNA targets of miR-129-3p, miR-146b-5p, miR-543 and miR-550a showed spatio-temporal regulation in the developing human brain [31]. Fourteen such transcripts were identified (rfx7, syt14, nova1, kiaa1199, spata13, osbpl8, exph5, ngef, pag1, acyp1, pde5a, adamts11, cyp26b1 and pag1) (Supplemental Material 6), suggesting that these targets should be the primary focus of further follow-up examinations in the brain tissue of neurodevelopmental disorders.

CONCLUSIONS

This study established the mRNA and miRNA expression changes in human fibroblast cell cultures in response to two nutrition-based metabolic stress paradigms. We found that 1) glucose depleted, galactose supplemented and a lipid reduced, cholesterol depleted medium produced widespread mRNA and miRNA changes in human fibroblasts; 2) while the most significantly changed miRNAs showed variation, the overall miRNA changes across the two metabolic stress conditions showed considerable similarity; 3) it appears that changes in miRNA expression contributed considerably to altered mRNA expression profile in both the GAL and RL conditions and 4) we identified 4 significantly changed miRNAs (miR-129-3p, miR-146b-5p, miR-543 and miR-550a) that are likely to be important mediators of multiple types of metabolic stress.

Using the individual transcript assessment approach, we identified only four significantly changed miRNAs (miR-129-3p, miR-146b-5p, miR-543 and miR-550a) across the two metabolic stressors. Yet, this is not surprising, as transcripts work in networks: while we observed considerable differences in the most changed miRNAs of the GAL- and RL-treated fibroblasts, the overall miRNA profile of the two stressors was strikingly similar. Furthermore, 4 metabolic stress-related transcript pathways (PPARA, CHREBP2, RACCYCD, and HSP27) were upregulated in both GAL and RL-exposed cultures, suggesting that this common signature is present both at the miRNA and mRNA level. This came as a surprise, as the two stressors acts through distinct mechanisms: the GAL condition primarily creates oxidative radicals through starvation [7], while the RL conditions forces the cells into a metabolic overdrive [10]. Based on these findings we propose that metabolic stress, regardless of the type of the insult, might have a common miRNome and mRNA signature, and this common signature perhaps represents a non-specific resilience response, helping the cells to adapt and survive.

Our results also suggest that the four changed miRNAs alone appeared to be able to regulate the mRNA expression of 52 target genes, arguing that miRNA regulation of the transcriptome might be a significant contributor to metabolic stress-induced cellular responses. Unfortunately, miR-543 and miR-550a has not been well studied experimentally, and their effects remain mostly unknown. However, mir-129-2 and miR-146b-5p have been a focus of multiple investigations to date: mir-129-2 is located next to a CpG island and is methylation-associated, and the expression of this miRNA can be epigenetically restored by de-methylating drugs [22]. Furthermore, mir-129-2 is conserved in vertebrates and regulates the expression of CP110, ARP2, TOCA1, ABLIM1 and ABLIM3, thus effectively controlling actin dynamics in the cell [21]. Still, miR-146b-5p might be the most important finding of our studies: its expression is highly induced by proinflammatory cytokines, with a clear dependency on IFN-γ, potentially via the JAK/STAT pathway [24,25]. Furthermore, a decrease of miR-146b-5p has been observed in monocytes during obesity, and it was associated with loss of the anti-inflammatory signaling action [26]. miR-146b-5p also regulates signal transduction of TGF- β [27], and appears to play a critical role in the pathophysiology of various cancers [28,29]. These data suggest that metabolic stress might regulate miRNA expression through epigenetic modification, and that the two metabolic stressors trigger an inflammation-like response in the stressed cells. However, this hypothesis has to be tested more comprehensively in follow-up experiments.

Finally, we need to emphasize that peripheral biomarkers studies are essential to understand the pathophysiology of human brain disorders. Recent studies indicate that peripheral biomarkers may aid the diagnosis of psychiatric disorders and help assess treatment efficacy [30]. However, obtaining peripheral patient tissues and analyzing them at resting state might not be sufficient to reveal the critical deficits, and they must be challenged to reveal the full signature of the disease [7,31]. Human dermal fibroblasts as a model system are ideally suited for such studies [32], as they have several advantages over the functional assays performed on peripheral blood cells: 1) they can be readily propagated without immortalization, 2) have a more similar expression profile to brain cells than peripheral blood cells; 3) are homogenous and very responsive to a variety of stimuli, and 4) they are minimally affected by epigenetic changes, which usually disappear after the 5th culture passage.

Our results also suggest that metabolic stress in dermal fibroblasts affects genes that are important for brain development and homeostasis. The 4 miRNAs with differential expression in both the RL and GAL condition (miR-129-3p, miR-146b-5p, miR-543 and miR-550a) regulated multiple genes (e.g. NGEF, NOVA1, PDE5A) that show region- and age-specific transcription in the human brain [33], suggesting that deregulation of these transcripts might have significant consequences on brain function, and should be examined in the context of neurodevelopmental disorders. Thus, dermal fibroblasts are appealing, readily available human biomaterials, allowing to perform controlled, functional and reproducible experiments on human tissue. This is especially important for studying major, polygenic psychiatric disorders, where postmortem tissue availability is limited and the genetic animal models do not fully recapitulate the human disease. This study defined the miRNA/mRNA transcriptome responses of fibroblasts to metabolic stress in healthy individuals, and similar studies should be performed in diseased patient material – the response of their transcriptome to the metabolic stress will be more revealing of pathophysiology than the case-control differences under resting, unchallenged conditions.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Acknowledgments

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Research highlights

- metabolic stress produced strong mRNA and miRNA changes in human fibroblasts
- miRNA changes across two metabolic stress conditions showed substantial similarity
- miRNA expression contributed to altered mRNA expression
- miRNA and mRNA changes were correlated
- miR 129 3p, miR 146b 5p, miR 543 and miR 550a are mediators of metabolic stress



Figure 1. Experimental design.

Flowchart representing the sample collection and grouping, the experiments and analyses performed. In summary, skin biopsies were collected from 17 male and female individuals, and used to establish *in vitro* fibroblast cultures. Each fibroblast culture was then cultured in standard (STD), galactose containing (GAL) or reduced lipid (RL) media. DNA microarrays were used to detect the mRNA profile of each individual sample in each culture condition. 4-5 samples of similar age and gender were grouped before performing miRNA expression profiling with miRNome qPCR arrays. Grouping-sample effects were examined in individual samples with custom qPCR arrays containing selected miRNA assays. Finally, we preformed correlation analyses between the commonly changed miRNAs and their putative mRNA targets.



Figure 2. qPCR validation of selected mRNA expression changes.

10 transcripts were validated using qPCR (SCD, FABP3, DHCR7, INSIG1, QPRT, HMGCR, HMGCS1, LDLR, FADS1, FADS2). X- axis denotes microarray-obtained average log2 ratios (ALR), while Y-axis denotes actin-normalized $-\Delta\Delta$ Ct values obtained by qPCR. Note that the microarray and qPCR data were highly correlated (GAL: r=0.92 p<0.001; RL: r=0.72, p<0.001). For further details, see Supplemental Material 4.



Figure 3. Commonly changed miRNAs in GAL and RL culture conditions.

45 and 34 miRNAs showed significant change in GAL and RL conditions, respectively (reddecreased; blue – increased). Note that the 4 commonly changed miRNAs had the same directionality in both GAL and RL conditions, and they were significantly altered in both pairwise (ppval) and groupwise (gpval) testing.



Figure 4. Overall miRNA pattern changes in GAL and RL conditions are highly correlated. The 45 miRNAs with significantly changed expression pattern in GAL conditions were assessed in the RL samples (**A**) and the 34 miRNAs with significantly changed expression pattern in RL conditions were assessed in the GAL samples (**B**). Note that the miRNA pattern observed in either metabolic stressor also showed a similar, highly correlated expression pattern in the other condition.



Figure 5. miRNA expression pattern separates the STD, GAL and RL lipid conditions.

Two-way unsupervised hierarchical clustering was performed on the expression level (Δ Ct) of 75 differentially expressed miRNAs. Samples were clustered vertically, miRNAs were clustered horizontally. Each colored square represents a normalized gene expression level, color coded for increase (red) or decrease (blue) from the mean. Color intensity is proportional to magnitude of change. The clustering resulted in a perfect separation of samples into three discrete groups corresponding to culture conditions (vertical dendrogram, green STD, yellow GAL, and orange RL).



Figure 6. Correlation of miRNA changes between individual and pooled fibroblast samples cultured in GAL (A) and RL (B) conditions.

22 miRNAs were assessed in the validation study, with the pooled $\Delta\Delta$ CTs plotted on X axis and the $\Delta\Delta$ CTs from individual samples plotted on the Y axis. Each miRNA represents a single data point. The correlation coefficient indicates that pooling of the samples did not skew artificially the miRNA pattern observed in GAL and RL.

Table 1

mRNA transcriptome changes. Microarray gene expression data, subjected to BioCarta pathway analysis with GSEA, identified 19 pathways enriched in fibroblasts cultured in GAL (**A**) and 15 in RL (**B**). The significantly changed BioCarta pathways are ranked by a net enrichment score (NES). The number of genes in each pathway is depicted in the Size column. Note that there are 4 differentially expressed pathways common for GAL and RL treatment (*).

A.	Galactose	(GAL)	media
	Gunactobe	(GILL)	meana

NAME	SIZE	NES	p-value
RACCYCD*	26	1.93	0.000000
CELLCYCLE	23	1.83	0.000000
SRCRPTP	11	1.70	0.000000
P53	16	1.70	0.002045
TEL	18	1.73	0.010730
STATHMIN	19	1.57	0.011516
CFTR	12	1.65	0.012000
CERAMIDE	22	1.67	0.012448
ATRBRCA	21	1.60	0.016129
G2	24	1.59	0.026157
HSP27 [*]	15	1.47	0.026639
CARM_ER	34	1.61	0.027451
PLCE	12	1.47	0.030864
CHREBP2*	42	1.61	0.031373
AKAPCENTROSOME	15	1.53	0.031809
CHEMICAL	22	1.56	0.032520
RB	13	1.53	0.034137
PPARA*	56	1.43	0.036000
ATM	20	1.58	0.044807

B. Reduced lipid (RL) media

NAME	SIZE	NES	p-value
PPARA*	56	1.80	0.000000
Р38МАРК	39	1.56	0.009579
VIP	26	1.60	0.009709
BAD	26	1.48	0.015936
CHREBP2*	42	1.64	0.022945
RACCYCD*	26	1.65	0.027559
GCR	19	1.46	0.038000
HSP27 [*]	15	1.49	0.040541

NAME	SIZE	NES	p-value
CD40	15	1.58	0.044000
LYM	11	1.37	0.047714
TNFR2	18	1.53	0.048583
TALL1	15	1.52	0.048638
IL22BP	16	-1.56	0.020638
DNAFRAGMENT	10	-1.51	0.030488
AHSP	11	-1.59	0.043222

*Commonly upregulated in NL and GAL treatment

Table 2

miRNA expression profile changes in GAL (A) and RL (B) treatment in human fibroblasts. Blue - increased expression; Red – reduced expression; $\Delta\Delta$ Ct - normalized threshold cycle difference between experimental and control samples; gt - groupwise t-test; pt – pairwise t-test. The correlations of these data are graphically represented in Figure 3.

GAL-induced miRNA expression changes and their expression in RL

miRNA ID	ddCt (GAL-STD)	gt-test (GAL-STD)	pt-test (GAL-STD)	ddCt (RL-STD)	gt-test (RL-STD)	pt-test (RL-STD)
hsa-miR-378b	-2.065	0.006	0.003	-0.512	0.548	0.614
hsa-miR-1183	-1.875	0.023	0.064	-0.485	0.557	N/A
hsa-miR-429	-1.801	0.019	0.154	-2.877	N/A	N/A
hsa-miR-3682-3p	-1.704	0.031	0.178	-0.212	0.659	0.936
hsa-miR-187*	-1.683	0.034	0.074	-1.270	N/A	N/A
hsa-miR-4302	-1.604	0.019	0.036	1.038	0.143	0.067
hsa-miR-375	-1.522	0.149	0.041	-0.417	0.535	0.862
hsa-miR-3200-5p	-1.283	0.041	0.128	0.338	0.489	0.509
hsa-miR-3161	-1.269	0.017	0.002	-0.015	0.982	0.865
hsa-miR-4267	-1.154	0.006	0.015	-0.915	0.105	0.085
hsa-miR-378	-1.115	0.057	0.040	0.935	0.169	0.183
hsa-miR-3156-5p	-1.055	0.022	0.034	-0.148	0.723	0.870
hsa-miR-550a	-1.054	0.031	0.025	-1.236	0.025	0.039
hsa-miR-4257	-1.044	0.020	0.327	-0.972	0.227	0.436
hsa-miR-3159	-0.943	0.142	0.037	-3.222	0.299	0.325
hsa-miR-550a*	-0.941	0.042	0.133	-1.192	0.130	0.072
hsa-miR-720	-0.903	0.090	0.023	-0.236	0.588	0.335
hsa-miR-2278	-0.839	0.287	0.035	-1.490	0.155	0.169
hsa-miR-3907	-0.827	0.070	0.013	0.291	0.436	0.343
hsa-miR-3687	-0.717	0.042	0.011	-0.370	0.567	0.648
hsa-miR-532-3p	-0.709	0.062	0.029	-0.069	0.820	0.757
hsa-miR-3190	-0.658	0.430	0.050	-0.930	0.242	0.572
hsa-miR-1247	-0.645	0.450	0.022	0.177	0.803	0.800
hsa-miR-3647-3p	-0.590	0.011	0.048	-0.017	0.949	0.960
hsa-miR-301a	0.602	0.038	0.025	0.802	0.070	0.060
hsa-miR-369-3p	0.612	0.029	0.030	0.575	0.052	0.025
hsa-miR-874	0.616	0.313	0.002	0.632	0.398	0.277
hsa-miR-486-5p	0.626	0.279	0.029	-0.196	0.754	0.680
hsa-miR-543	0.659	0.025	0.020	0.610	0.021	0.035
hsa-miR-1260b	0.660	0.077	0.049	0.290	0.459	0.431
hsa-miR-424	0.698	0.061	0.011	0.919	0.104	0.052
hsa-miR-127-5p	0.713	0.069	0.045	0.696	0.120	0.114
hsa-miR-3913-5p	0.757	0.343	0.031	N/A	N/A	N/A

miRNA ID	ddCt (GAL-STD)	gt-test (GAL-STD)	pt-test (GAL-STD)	ddCt (RL-STD)	gt-test (RL-STD)	pt-test (RL-STD)
hsa-miR-628-5p	0.871	0.020	0.067	0.355	0.440	0.552
hsa-miR-181a*	0.876	0.039	0.014	0.759	0.213	0.146
hsa-miR-129-3p	0.949	0.011	0.053	1.108	0.020	0.051
hsa-miR-146b-5p	1.058	0.007	0.011	1.487	0.015	0.005
hsa-miR-1271	1.227	0.032	0.113	1.338	0.171	0.339
hsa-miR-1285	1.386	0.010	0.037	0.667	0.407	0.517
hsa-miR-337-5p	1.550	0.017	0.103	1.200	0.060	0.061
hsa-miR-217	2.018	0.112	0.013	1.323	0.097	0.223
hsa-miR-3176	2.041	0.018	0.009	0.635	0.373	0.515
hsa-miR-497*	2.313	0.040	0.079	1.803	0.098	0.204
hsa-miR-1227	2.462	0.014	0.113	0.488	0.343	0.172
hsa-miR-146b-3p	3.795	0.046	0.123	3.291	0.216	0.205

RL-induced miRNA expression changes and their expression in GAL

miRNA ID	ddCt (RL-STD)	gt-test (RL-STD)	pt-test (RL-STD)	ddCt (GAL-STD)	gt-test (GAL-STD)	pt-test (GAL-STD)
hsa-miR-3154	-1.342	0.039	0.047	-2.425	0.161	0.267
hsa-miR-550a	-1.236	0.025	0.039	-1.054	0.031	0.025
hsa-miR-16-1*	-1.181	0.014	0.063	-0.437	0.347	0.311
hsa-miR-138-1*	-0.924	0.326	0.021	0.219	0.775	0.712
hsa-miR-641	-0.887	0.042	0.253	-0.059	0.875	0.726
hsa-miR-2110	-0.805	0.043	0.031	0.017	0.953	0.965
hsa-miR-320e	-0.618	0.032	0.118	0.162	0.384	0.513
hsa-miR-376c	0.493	0.039	0.027	0.490	0.114	0.050
hsa-miR-296-5p	0.561	0.011	0.018	0.085	0.736	0.623
hsa-miR-517*	0.595	0.099	0.013	0.245	0.475	0.375
hsa-miR-543	0.610	0.021	0.035	0.659	0.025	0.020
hsa-miR-214	0.636	0.006	0.032	0.703	0.088	0.075
hsa-miR-195	0.689	0.011	0.061	0.163	0.600	0.586
hsa-miR-18a*	0.692	0.057	0.014	-0.504	0.238	0.141
hsa-miR-19a	0.710	0.103	0.035	0.168	0.660	0.323
hsa-miR-212	0.742	0.031	0.032	0.404	0.155	0.078
hsa-miR-132*	0.753	0.120	0.002	0.766	0.122	0.292
hsa-miR-411	0.779	0.016	0.066	0.331	0.391	0.410
hsa-miR-330-5p	0.854	0.117	0.007	0.797	0.261	0.347
hsa-miR-431	0.864	0.023	0.004	0.579	0.063	0.016
hsa-miR-154*	0.891	0.008	0.115	1.052	0.087	0.110
hsa-miR-18a	0.900	0.039	0.029	0.198	0.597	0.375
hsa-miR-29a*	0.912	0.041	0.116	0.801	0.085	0.220
hsa-miR-135b	0.927	0.148	0.037	0.698	0.395	0.249
hsa-miR-22*	1.033	0.049	0.194	0.966	0.083	0.339

miRNA ID	ddCt (RL-STD)	gt-test (RL-STD)	pt-test (RL-STD)	ddCt (GAL-STD)	gt-test (GAL-STD)	pt-test (GAL-STD)
hsa-miR-31	1.105	0.004	0.012	0.750	0.065	0.079
hsa-miR-129-3p	1.108	0.020	0.051	0.949	0.011	0.053
hsa-miR-758	1.188	0.042	0.042	0.199	0.672	0.608
hsa-miR-106b*	1.193	0.035	0.229	0.188	0.795	0.918
hsa-miR-299-3p	1.414	0.048	0.109	0.612	0.159	0.243
hsa-miR-146b-5p	1.487	0.015	0.005	1.058	0.007	0.011
hsa-miR-196b*	1.599	0.025	0.034	1.226	0.114	0.289
hsa-miR-21*	1.647	0.003	0.019	1.102	0.093	0.052
hsa-miR-331-5p	2.660	0.010	0.041	0.058	0.915	0.932

Table 3

miRNA and mRNA expression changes are correlated in human fibroblasts. The three commonly changed miRNAs in both GAL and RL conditions were correlated with expression changes of 52 putative target mRNAs. **A**. miR-129-3p target changes, **B**. miR-550a target changes, **C**. miR-146b-5p target changes. The commonly changed mRNAs in both GAL- and RL-medium exposed human fibroblasts are marked with an asterisk.

A. miR-129-3p

Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
FADS1*	1.08	0.000001	FADS1*	1.22	0.000000
FADS1*	1.08	0.000001	FADS1*	1.22	0.000000
FADS1*	1.06	0.000001	FADS1*	1.11	0.000000
DGKI	0.68	0.000073	STARD4	0.94	0.000000
LHFPL2	0.63	0.000000	TMTC1*	0.57	0.000173
PTGER4*	0.51	0.000302	PTGER4*	0.57	0.000329
TMTC1*	0.50	0.000007	CCDC102B	0.51	0.006941
CDK1	0.49	0.007441	RCAN2	0.45	0.000727
CDCA7L	0.47	0.000002	CHML	0.43	0.000010
PKIA	0.43	0.000075	RFX7	0.41	0.000179
C2orf88	0.41	0.000036	RIMS1	0.38	0.004726
RDH10	-0.42	0.000011	CYP27C1	0.38	0.000169
			ZNF436	-0.42	0.000329
			BAALC	-0.60	0.000524

B. miR-550a

Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
CLIC2	1.18	0.000000	DSTN	0.44	0.000322
SPATA13	0.50	0.000059	EXPH5*	0.41	0.008081
C1QTNF2	0.48	0.002883	NGEF	0.40	0.005541
OSBPL8	0.48	0.000000	APOL6	-0.93	0.000000
PLD1	0.45	0.000245			
SCARA3	0.39	0.000014			
ADAMTS6	-0.38	0.005308			
EXPH5*	-0.51	0.001573			
MYH2	-0.80	0.000192			

C. miR-146b-5p

Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
IFIT3	0.65	0.000003	$SVIL^*$	0.58	0.000000
IQGAP3	0.64	0.000058	KRTAP1-5*	0.56	0.000110

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Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
IRAK1	0.61	0.000001	${ m SVIL}^*$	0.50	0.000009
SYT14	0.61	0.000101	CLCN6*	0.46	0.000000
RAB7L1	0.58	0.000000	PID1	0.42	0.000173
RAB7L1	0.55	0.000000	NOVA1	0.41	0.000848
CLCN6*	0.52	0.000001	SLC1A1*	-0.45	0.000093
SVIL*	0.44	0.000079	LRRC15	-0.45	0.000372
STRBP	0.43	0.000607	KIAA1199	-0.47	0.001176
TIMELESS	0.42	0.000047			
RACGAP1	0.41	0.000632			
STIL	0.40	0.002208			
WISP1	-0.38	0.002183			
DNAL1	-0.39	0.000081			
DNAL1	-0.40	0.002178			
SLC1A1*	-0.49	0.000259			
KRTAP1-5*	-0.53	0.003753			
RASGRP1	-0.67	0.000057			

D. miR-543

Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
RRAGD*	2.36	0.000000	NRXN3	1.98	0.009901
RRAGD	1.48	0.000000	ABCA1*	1.60	0.000000
PAG1	0.98	0.000000	ABCA1*	0.92	0.000000
PAG1	0.91	0.000000	PLA2G4A	0.86	0.002028
LAMA1	0.83	0.000029	GALNT1	0.84	0.003950
C3orf55	0.82	0.000002	RRAGD*	0.79	0.009357
PAG1*	0.80	0.000000	PPP1R12B*	0.76	0.003190
PRKAR2B*	0.76	0.000002	GALNT1	0.70	0.005351
TOX	0.68	0.000073	GALNT1	0.63	0.002321
ZNF367	0.64	0.000001	COL16A1	0.61	0.000219
CDK1	0.63	0.000099	ZNF436	0.56	0.000329
PBX3	0.62	0.000000	TRIM13	0.50	0.000035
DCLK1	0.61	0.000864	ZNF436	0.49	0.001606
FNIP2*	0.58	0.000013	DUSP3*	0.46	0.000321
ZCCHC14	0.58	0.000000	EVI5	0.39	0.000685
CDK1	0.58	0.000135	TGFB1	0.39	0.000077
GTSE1	0.56	0.001911	DDX3X	0.38	0.005765
CDK1	0.56	0.000136	PAG1*	0.38	0.000839
BEX2	0.55	0.000005	FNIP2*	0.38	0.000458
GTSE1	0.54	0.001029	CCDC117	-0.38	0.000395
DUSP3*	0.54	0.000000	ADAMTSL1	-0.41	0.000269

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	Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
ĺ	FNIP2	0.52	0.000006	EEF1A1	-0.42	0.003515
	ESCO2	0.51	0.000066	SMAD5	-0.42	0.008163
	E2F7	0.51	0.000144	ADAMTSL1*	-0.43	0.002492
	FMNL2	0.49	0.000014	TMEM133	-0.43	0.003737
	B3GNT5	0.49	0.000143	ADAMTSL1*	-0.44	0.004797
	ABCA1*	0.49	0.000012	PDP1	-0.44	0.000010
	CDK1	0.49	0.007441	ADAMTSL1	-0.53	0.000006
	OSBPL8	0.48	0.000000	CHML	-0.53	0.000279
	DCLK1	0.48	0.003116	PDE5A*	-0.56	0.002906
	DCLK1	0.48	0.003116	GBP1	-0.60	0.003012
	SLC2A13	0.47	0.000000	RAPH1*	-0.68	0.000064
	OSBPL1A	0.47	0.000000	HIPK2	-0.71	0.00009
		0.47	0.000006	DDK 1 D 2 D *	-0.75	0.000159
		0.46	0.000304	CDED5*	1 40	0.000041
	RADJIAH	0.40	0.000394	CREB5	-1.49	0.000041
	DUSP3	0.46	0.000000	CREB5	-1.58	0.000760
	LAMP2	0.45	0.000000			
	GISEI	0.45	0.003273			
	FMNL2	0.45	0.000000			
	WHSCI	0.44	0.000477			
	STRBP	0.43	0.000607			
	SMC4	0.43	0.000303			
	DUSP2	0.42	0.000009			
	C3orf55	0.42	0.000000			
	SI C2A13	0.41	0.000024			
	ATP6V1A	0.40	0.0000000			
	OSBPL8	0.40	0.000000			
	HMGB2	0.40	0.000029			
	ACYP1	0.40	0.000001			
	RBM47	0.40	0.000154			
	MEGF9	0.40	0.000002			
	ZCCHC14	0.40	0.000657			
	OSBPL1A	0.40	0.000003			
	CD163	0.39	0.005055			
	ATP6V1A	0.39	0.000022			
	WHSC1	0.38	0.000535			
	TMEM135	0.38	0.000045			
	SC5DL	0.38	0.000265			
	NRXN3	-0.38	0.002052			
	USP53	-0.38	0.002009			
	NRCAM	-0.38	0.000730			
	PDE5A*	-0.39	0.000332			

Gene	ALR GAL	pval GAL	Gene	ALR RL	pval RL
MDM4	-0.40	0.000436			
COL1A2	-0.41	0.000018			
TRIM13	-0.41	0.001256			
SNAPC3	-0.43	0.000598			
ADAMTSL1*	-0.43	0.000002			
ARHGAP29	-0.45	0.000223			
RAPH1	-0.45	0.000001			
SYNE1	-0.47	0.000020			
GATS	-0.47	0.000009			
RAPH1*	-0.49	0.000071			
CREB5*	-0.50	0.000029			
RAPH1	-0.52	0.000001			
RAPH1	-0.52	0.000007			
ACVR2A	-0.52	0.000006			
CHN1	-0.53	0.000002			
FNDC3B	-0.54	0.000034			
CACNB4	-0.56	0.000001			
SYTL4	-0.56	0.000113			
ADAMTSL1*	-0.57	0.000000			
CREB5*	-0.59	0.000234			
ADAMTSL1	-0.60	0.000003			
PPP1R12B*	-0.63	0.000179			
PAWR	-0.64	0.000017			
ID4	-0.74	0.000000			
ID4	-0.80	0.000000			
ID4	-0.83	0.000002			
ARRDC4	-0.85	0.000002			
CYP26B1	-1.16	0.000000			

L ALR- average log2 ratio; RL- reduced lipid media; GAL- galactose media

* commonly changed mRNAs in GAL and NL treatment