

SHORT REPORT

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Transcriptional regulatory program in wild-type and retinoblastoma gene-deficient mouse embryonic fibroblasts during adipocyte differentiation

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Abstract

Background: Although many molecular regulators of adipogenesis have been identified a comprehensive catalogue of components is still missing. Recent studies showed that the retinoblastoma protein (pRb) was expressed in the cell cycle and late cellular differentiation phase during adipogenesis. To investigate this dual role of pRb in the early and late stages of adipogenesis we used microarrays to perform a comprehensive systems-level analysis of the common transcriptional program of the classic 3T3-L1 preadipocyte cell line, wild-type mouse embryonic fibroblasts (MEFs), and retinoblastoma gene-deficient MEFs (Rb-/ MEFs).

Findings: Comparative analysis of the expression profiles of 3T3-L1 cells and wild-type MEFs revealed genes involved specifically in early regulation of the adipocyte differentiation as well as secreted factors and signaling molecules regulating the later phase of differentiation. In an attempt to identify transcription factors regulating adipogenesis, bioinformatics analysis of the promoters of coordinately and highly expressed genes was performed. We were able to identify a number of high-confidence target genes for follow-up experimental studies. Additionally, combination of experimental data and computational analyses pinpointed a feedback-loop between Pparg and Foxo1.

To analyze the effects of the retinoblastoma protein at the transcriptional level we chose a perturbated system (Rb-/ MEFs) for comparison to the transcriptional program of wild-type MEFs. Gene ontology analysis of 64 deregulated genes showed that the Rb-/ MEF model exhibits a brown(-like) adipocyte phenotype. Additionally, the analysis results indicate a different or additional role for pRb family member involvement in the lineage commitment.

Conclusion: In this study a number of commonly modulated genes during adipogenesis in 3T3-L1 cells and MEFs, potential transcriptional regulation mechanisms, and differentially regulated targets during adipocyte differentiation of Rb-/ MEFs could be identified. These data and the analysis provide a starting point for further experimental studies to identify target genes for pharmacological intervention and ultimately remodeling of white adipose tissue into brown adipose tissue.

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Introduction

Evidence accumulating during the past decades has convincingly revealed that adipocytes and adipose tissue are not only acting as a storage depot for fat, but are also actively involved in regulating whole body energy balance, and much attention has been dedicated to decipher molecular events and to identify factors involved in fat cell development. Although many molecular regulators have been identified, a comprehensive catalogue of components regulating adipogenesis is still missing. Therefore, a number of expression profiling studies using microarrays have been performed to monitor global gene expression profiles during *in vitro* adipocyte differentiation of several cell models [1].

Recently, we have studied global gene expression of 3T3-L1 cells during adipogenesis and identifying molecular mechanisms, biological processes, key enzymes, including many characterized and uncharacterized transcriptional regulators, thereby providing a gene expression atlas using functional annotation [2]. We showed that the mitotic clonal expansion phase plays an important role during adipocyte differentiation in the studied mouse cell model. One group of proteins - the pocket proteins including the retinoblastoma protein (pRb) and other members (p130, p107) - was expressed during both, the cell cycle phase and late cellular differentiation. This dual role of pRb in the early and late stages of adipogenesis was also described in other studies [3-6]. Pathways regulated by the retinoblastoma gene/protein might therefore be interesting novel targets in the combat to curb or to treat obesity and associated disorders.

Several studies on mechanisms and interactions have elucidated how pRb - the product of the Rb1 gene - fulfills its functions. pRb is phosphorylated by cyclin-dependent kinases thereby regulating E2F activity [5]. Other mechanisms include co-factor activity for non-E2F transcription factors, recruitment of chromatin remodeling complexes (including HDACs) and action as a non-chromatin associated protein adaptor [4,6,7]. The first evidence that pocket proteins are involved in adipocyte differentiation emerged by showing interactions with the C/EBP transcription factors [8,9]. It was shown that the ability of the C-terminally truncated large T antigen (TAg) to inhibit differentiation of white preadipocyte cell lines is dependent on the ability to sequester pRb family members [10]. Further studies showed that pRb and p107/p130 may have opposite effects on adipocyte differentiation [11,12].

Rb-/- mouse embryonic fibroblasts (MEFs) are unable to undergo adipose conversion using standard adipogenic inducers [8,13,14]. It was suggested that the underlying mechanism did not directly mirror cell cycle-related aspects of pRb function, as Rb-/- cells enter and exit mitotic clonal expansion with the same kinetics as wild-type cells [13]. Further it was shown that pRb is

required for preadipocyte differentiation *in vivo* [12,15] and functions as a molecular switch determining white versus brown(-like) adipocyte differentiation [14,16]. Of note, adipose tissue-specific inactivation of Rb prevents [15] and haploinsufficiency of Rb attenuates diet induced obesity and insulin resistance [17]. pRb might also be involved in the lineage commitment to osteogenic or adipogenic differentiation of mesenchymal stem cells [18].

Thus, given the accumulated evidence of the involvement of pRb in adipogenesis we initiated a large scale gene expression study to elucidate its role during adipogenic differentiation. We first performed microarray gene expression analysis of wild-type mouse embryonic fibroblasts (MEFs) undergoing adipogenesis and compared these expression data with microarray gene expression data from 3T3-L1 cell adipogenesis to identify a transcriptional program common between the two cell models. Additionally, bioinformatics analyses were carried on the promoters of highly expressed genes to identify potential transcription factor binding sites and to generate hypotheses on transcription factor binding. To analyze the effects of the retinoblastoma protein at the transcriptional level we chose a perturbated system (Rb-/- MEFs) for comparison to the transcriptional program of wild-type MEFs.

Results and Discussion

Common transcriptional program of adipocyte differentiation in MEFs and 3T3-L1 cells

To compare transcriptional regulation during adipocyte differentiation of the established 3T3-L1 cell line and mouse embryonic fibroblasts (MEFs), gene expression in MEFs was studied in three independent experiments with whole genome cDNA microarrays (>27k elements) using seven time points (d0, d1, d2, d3, d4, d6, d10) after hormonal induction in relation to expression levels at the preconfluent stage (similar to the generation of data from the previous 3T3-L1 study [2]) (Additional file 1). We validated the microarray data from MEFs by quantitative reverse transcription polymerase chain reaction (qPCR) analysis of the first differentiation experiment (see Figure S1 in Additional file 2) with six genes at several time points. Applying linear regression of the resulting 18 data points showed a coefficient of determination (R^2) of 0.92 (Pearson correlation coefficient $r = 0.96$), which represented good agreement between microarray and qPCR results. Replicate results were also in good concordance with log₂-fold changes averaged over the three independent differentiation experiments. Differentially expressed genes with similar combined expression profiles of the 3T3-L1 and MEF cells were grouped into 6 clusters by k-means clustering (Figure 1 and Additional file 1). Proficient quality of clustering

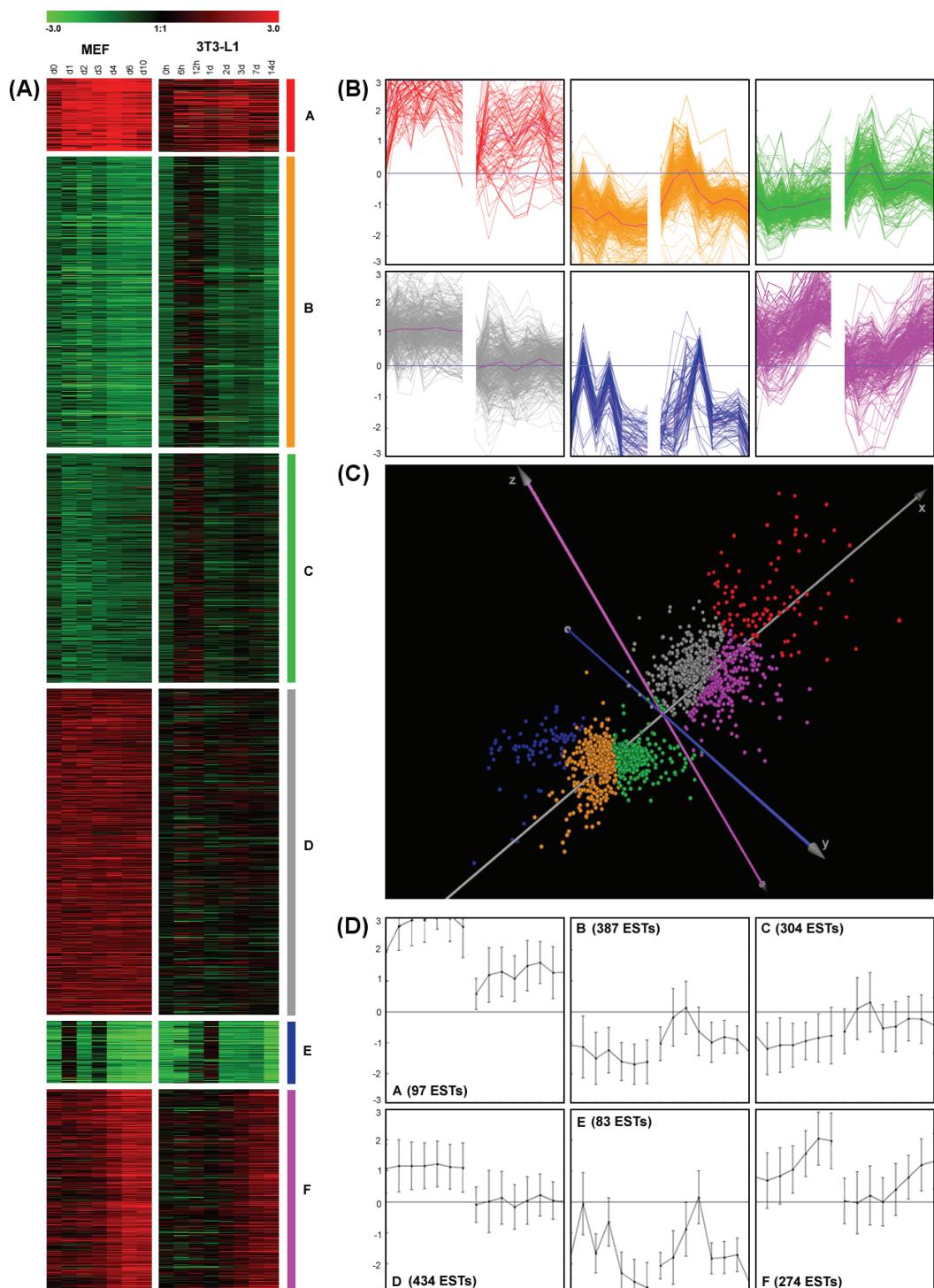


Figure 1 Gene expression levels (log₂-fold change) of genes from different time points during adipocyte differentiation in 3T3-L1 and MEF cells. **(A)** Compact heatmaps from k-means clustering. **(B)** Line drawing centroid view (mean +/- SD). **(C)** Three-dimensional visualization of Principal Component Analysis (PCA) (color coding according k-means clustering) demonstrates cluster separation. **(D)** Line drawing for expression course of each gene individually.

was evident from different visualizations of the gene expression levels and separation of clusters using Principal Component Analysis (PCA) in three-dimensional view of the first three principal components and color-coding according to k-means clustering (Figure 1).

Functional assignment of genes within each cluster by gene ontology analysis revealed a number of noticeable biological processes regulated at the transcriptional level (Figure 2). Most significantly overrepresented biological processes were 'cell cycle', 'cell division', 'DNA replication', and 'mitosis' indicating mainly the involvement of genes of cluster E and cluster B in cell cycle processes. Distinct peaks in the expression profiles of these genes at d1 and d3 showed that MEFs underwent two cycles of mitotic clonal expansion comparable to the 3T3-L1 cells showing - at the used sampling rate - one peak in the expression profiles at 24 h [2] which could reflect the first of two rounds of mitotic clonal expansions [19,20]. This observation is supported by reported phosphorylation events of the pRb in 3T3-L1 cells during differentiation [3,13] and the cyclin A2 (Ccna2) expression profile as shown in Additional file 1 and Additional file 2.

As evident from the gene ontology analysis of genes in cluster E, significantly overrepresented biological processes were 'fatty acid metabolism', 'lipid metabolism', 'metabolism', and 'fatty acid beta-oxidation'. Well known genes important for the development of the adipocyte phenotype were combined in cluster F with increasing expression over time, including transcription factors and nuclear receptors Pparg, Cebpa, Srebf1, Xbp1, Thra, Stat1, Nr0b2 (SHP), Nr1d1 (REV-ERB alpha), Nr3c1 (GR), Foxc1, Foxo1, Zhx1, and Zfp503. The recently identified gene coding for adipocyte plasma membrane-associated protein [21] was also shown to be highly expressed at late stages of adipogenesis of MEFs (cluster F).

Highly expressed genes over the differentiation course, especially at the middle stage of adipogenesis (after mitotic clonal expansion) could be found in cluster A. In this group we observed not only transcription factors like Klf9 [22], and Foxo3a [23] but also secreted factors (as studied in human [24]) including Apod [25], Ccdc80 [26], Adamts5 [27], Mmp2 [28], and Ptn [29], as well as enzymes, signaling molecules, and other factors such as Xdh [30], Enpp2 [31], Rgs2 [32], Txnip [33], Tsc22d3 [34], H6pd [35], Igfbp2 [36], and Igfbp3 [37]. Hence, the genes in cluster A represent potential novel players in adipogenesis and are therefore candidates to be studied in more detail experimentally.

To identify candidates selectively associated with the MEF cell model, genes upregulated during MEF differentiation which showed marginal changes in 3T3-L1 adipocytes compared to the preconfluent stage of 3T3-

L1 are of major interest (cluster D). Genes in this cluster were overrepresented in the GO terms 'heart development', 'negative regulation of cell proliferation', 'cell adhesion', 'extra cellular matrix organization and biogenesis', and 'positive regulation of angiogenesis'. The expressed transcription factors include Arnt, Arnt2, Atf5, Ets2, Egr1, Fosl2, JunB, Nfia, Prdm4, Stat2, Stat3, Tef, and Tcf20 which might be involved specifically in early regulation of adipocyte differentiation in the primary cell model as opposed to the already committed 3T3-L1 preadipocytes.

Transcription factors regulating gene expression during adipocyte differentiation

The terminal development of a fat cell is driven by a transcriptional cascade governed mainly by Pparg and Cebpa (as reviewed in [38,39]). The transcriptional events directing early adipogenic processes, concerted by molecular events prior to the activation of Pparg and Cebpa have not been exhaustively studied. Hence, we asked if there are common regulators of expression of genes within each cluster, especially those with expression profiles similar to a transcription factor and its target genes. To address this question, we calculated overrepresentation of sequence motifs (response elements) within a genomic region (including promoter) from -3 kb to +2 kb relative to transcription start site (TSS) of those genes (Additional file 3).

From the transcription factors with exclusively overrepresented binding sites (Figure 3A) in promoters of genes from cluster F (i.e. genes with increasing expression towards terminal adipocyte differentiation) only Foxo1 and the key player in adipogenesis Pparg have expression profiles in cluster F. It has been previously shown that Foxo1 can interact with Pparg [40]. Furthermore, Foxo1 can bind to the Pparg promoter region and suppress Pparg expression [41]. Our analyses suggest that a feed-back loop from Pparg to Foxo1 might exist. To provide further evidence we analyzed two datasets of DNA binding by Pparg identified by chromatin immunoprecipitation (ChIP) followed by sequencing [42] or by microarray analysis [43] during 3T3-L1 adipocyte differentiation. Integrative analysis of these datasets and Pparg binding sites in genomic regions around genes with expression profile similar to that of Pparg in both the 3T3-L1 cells and MEFs (cluster F), revealed that Foxo1 might indeed be regulated by Pparg as indicated in Figure 3B. Thus, our data, experimental results from other studies, and further computational analysis pinpointed a feed-back loop between Pparg and Foxo1.

We have recently identified 2310001A20Rik (Apmap), as well as the retrotransposed genes 6530401D17Rik (Arxes1) and 2900062L11Rik (Arxes2) as Pparg targets and that they are required for adipogenesis [21,44].

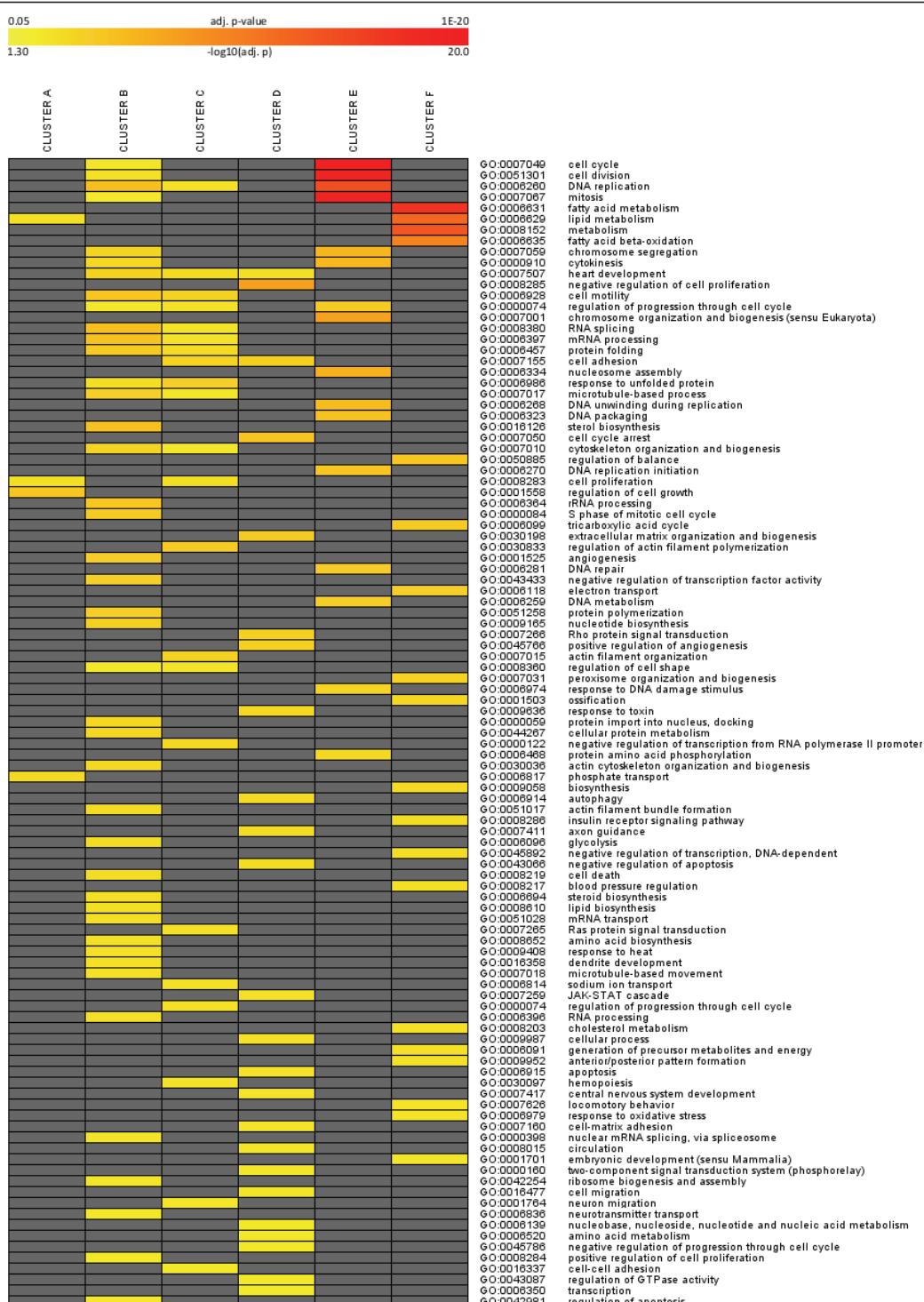


Figure 2 Gene ontology analysis. Heatmap for overrepresented gene ontology terms for genes in 6 clusters over several time points during adipocyte differentiation in 3T3-L1 and MEF cells as visualized in Figure 1 (color coding according to legend at the top, only gene ontology terms with adjusted $p < 0.05$ were considered) is shown.

(A)

Matrix ID	Sequence logo	Transcription factor	Sim	# targets (cluster F)	# genes (cluster F)	# targets (RefSeq)	# genes (RefSeq)	p-value	FDR
M00183		c-Myb	0.99	165	206	15707	21408	1.5E-02	1.8E-01
M00191		ER (estrogen receptor)	0.94	140	206	12979	21408	1.7E-02	1.9E-01
M00474		FOXO1 (fork head box O1)	0.86	193	206	18983	21408	1.1E-02	1.5E-01
M00477		FOXO3 (fork head box O3)	0.85	197	206	19614	21408	1.8E-02	1.9E-01
M00054		NF-kappaB	0.91	140	206	13100	21408	2.6E-02	2.2E-01
M00052		NF-kappaB (p65)	0.90	157	206	14819	21408	1.6E-02	1.8E-01
MA0061		NF-kappaB Vertebrates REL	0.92	126	206	11763	21408	4.1E-02	2.8E-01
M00424		NKX6-1	0.83	199	206	19985	21408	3.2E-02	2.5E-01
M00528		PPAR gamma	0.84	131	206	12135	21408	2.6E-02	2.2E-01
MG0001		PPARgamma/RXRalpha	0.84	123	206	10675	21408	2.7E-03	7.1E-02
ML0001		PPARgamma/RXRalpha	0.85	137	206	12750	21408	2.4E-02	2.2E-01
MA0065		PPARgamma-RXRalpha Homo sapiens	0.77	144	206	13391	21408	1.6E-02	1.8E-01
MA0071		RORalpha-1 Homo sapiens	0.96	168	206	15231	21408	3.9E-04	2.0E-02
M00156		RORalpha1	0.95	171	206	16108	21408	4.7E-03	9.8E-02
M00418		TGIF (5'TG3' interacting factor)	0.97	142	206	13264	21408	2.2E-02	2.1E-01

(B)

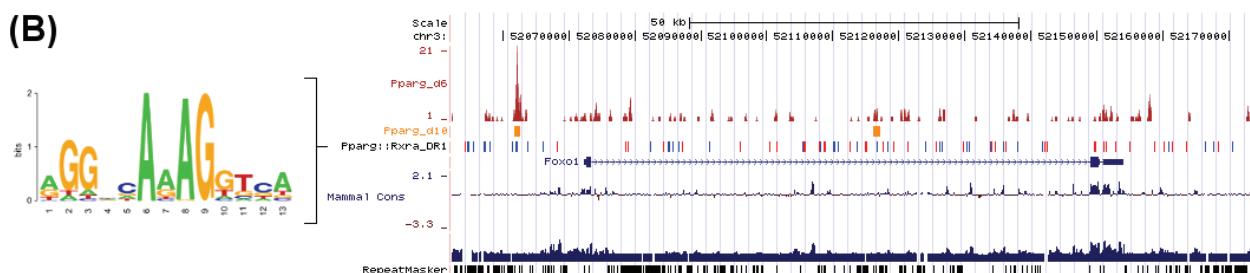


Figure 3 *In silico* promoter analysis and genomic organization. (A) Statistically overrepresented ($p < 0.05$) potential transcription factor binding sites (TFBS) identified by *in silico* analysis of genomic regions from -3 kb to +2 kb around transcription start sites of genes with increasing gene expression at late stages of adipocyte differentiation (cluster F) using position frequency (weight) matrices from TRANSFAC and JASPAR. Corresponding sequence logos are shown and number of targets within the dataset (cluster) is given and tested against the number of targets from all RefSeq transcripts using Fishers exact test. Note only transcription factor motifs are considered which are exclusively overrepresented in cluster F. (B) The genomic organization around Foxo1 using the UCSC genome browser (NCBI37/mm9) with high similarity to the Pparg-Rxra DR-1 motif (similarity score > 0.90) (blue '+' strand, red '-' strand), Pparg binding regions as identified by previous studies at day 6 [42] (dark red) and day 10 [43] (orange) of 3T3-L1 adipocyte differentiation.

A number of other genes annotated as Riken clones (like 5730469M10Rik and F630110N24Rik) included in cluster F show noticeable binding by Pparg in 3T3-L1 cells in the first intron and hence might be interesting targets in follow-up experimental studies. Additional target genes

were Acadm, Adcy5, Dgat1, Etfb, Ech1, Hmox1, Mdh1, Mgst3, Pparg, Pla2g15, Phldb1, Sorbs1, and Srebfb1 as indicated in both chromatin immunoprecipitation (ChIP) studies [42,43] of Pparg binding at the same genomic region near TSS (< 1 kb) or in an intron (peak

height > 70) (see additional website <http://icbi.at/adipo>). There was also supporting evidence of overrepresented known cell cycle related transcription factor binding sites in the genes of cluster B, C, and E (including NF-Y, E2F, cMyc:Max, USF, c-ETS). In cluster A exclusively overrepresented transcription factor binding sites includes STAT3, STAT5A, STAT5B, MEF-2, Hand1:E47, and NF-1. The genes encoding for STAT3 and NF-I could also be found in this cluster, indicating that these two transcription factors share similar expression profiles as their target genes in cluster A.

Transcriptional program of Rb-/ MEFs

Recent studies showed that the retinoblastoma protein (pRb) was expressed in the cell cycle and late cellular differentiation phase during adipogenesis. The interesting Rb1 gene expression profiles during 3T3-L1 and wild-type MEF adipogenesis are also shown in Additional file1. Consequently, we studied the effect of perturbation of the MEF adipogenesis system by lack of Rb1 by a direct comparison of the global expression levels of Rb-/ MEF (ME3) cells to those of the wild-type MEFs (MEFA) at 3 time points (d1, d3, d8) during adipocyte differentiation in relation to the expression at day 0, immediately before hormonal induction using again whole genome cDNA microarrays (> 27k elements). Since Rb-/ MEFs do not differentiate into adipocytes in response to a standard adipogenic cocktail, the proadipogenic Pparg agonist Rosiglitazone was added in both the Rb-/ MEF (ME3) and the Rb+/+ MEF (MEFA) differentiation experiments. We calculated several distance measures (sum of differences, Manhattan distance, Euclidean distance, and Pearson correlation) between the expression profiles of ME3 and MEFA cells for 3118 differentially expressed ESTs (Additional file 4) and focused on 64 upregulated genes (sum of differences > 4) and 90 downregulated genes (sum of differences < -4). For clarity, gene expression levels of only the most deregulated ESTs were visualized as heatmap in Figure 4 (23 ESTs with sum of differences > 5 and 41 ESTs with sum of differences < -5).

As evident from previous studies, microarray and qPCR analysis revealed that Ucp1, Ppargc1a, Prdm16, and Cidea are highly upregulated in Rb-/ MEF (ME3) adipocyte differentiation compared to Rb+/+ MEF (MEFA) adipocyte differentiation (see [14,16,45], Additional file 2, and Additional file 4). Gene ontology analysis from the 64 deregulated genes corroborated the brown(-like) adipocyte phenotype of the Rb-/ MEF model: the most significant overrepresented gene ontology term was 'generation of precursor metabolites and energy' (FDR = 3.1E-05) (including electron transferring flavoprotein, alpha polypeptide (Etfa) and NADH dehydrogenase (ubiquinone) Fe-S protein 1 (Ndufs1) also involved in

oxidative phosphorylation a prominent process in brown adipose tissue). Analysis with the tool ClueGO (Figure 5) showed prominent gene ontology terms 'mitochondrion organization', 'electron transport', 'triglyceride metabolic processes', and 'apoptosis'. These are related to hallmarks of a brown adipocyte phenotype such as mitochondrial biogenesis, metabolic processes, electron transport, oxidative phosphorylation, and uncoupling. Many of the activated genes are encoding for enzymes involved in glucose or lipid metabolism and one gene for a transcription factor (Myc).

It is noteworthy that gene set enrichment analyses of the ranked gene list with gene sets from brown adipose tissue versus white adipose tissue [46] and brown and white preadipocytes at the differentiating stages [47] showed a high enrichment score at least to the first gene set supporting the notion that Rb is involved in the switch between the white and brown adipocyte differentiation process but also indicated expression phenotypic differences between brown adipocytes and differentiated Rb-/ MEFs (Additional file 5). These results suggest a different and/or additional role of pRb. We have seen that Rb-/ MEFs behave more like primary cells from inguinal white adipose tissue than classic brown adipose tissue [48]. It was also shown that pRb could have a different role as it might be involved in the lineage commitment to either osteogenic or adipogenic differentiation of mesenchymal stem cells [18].

Conclusion

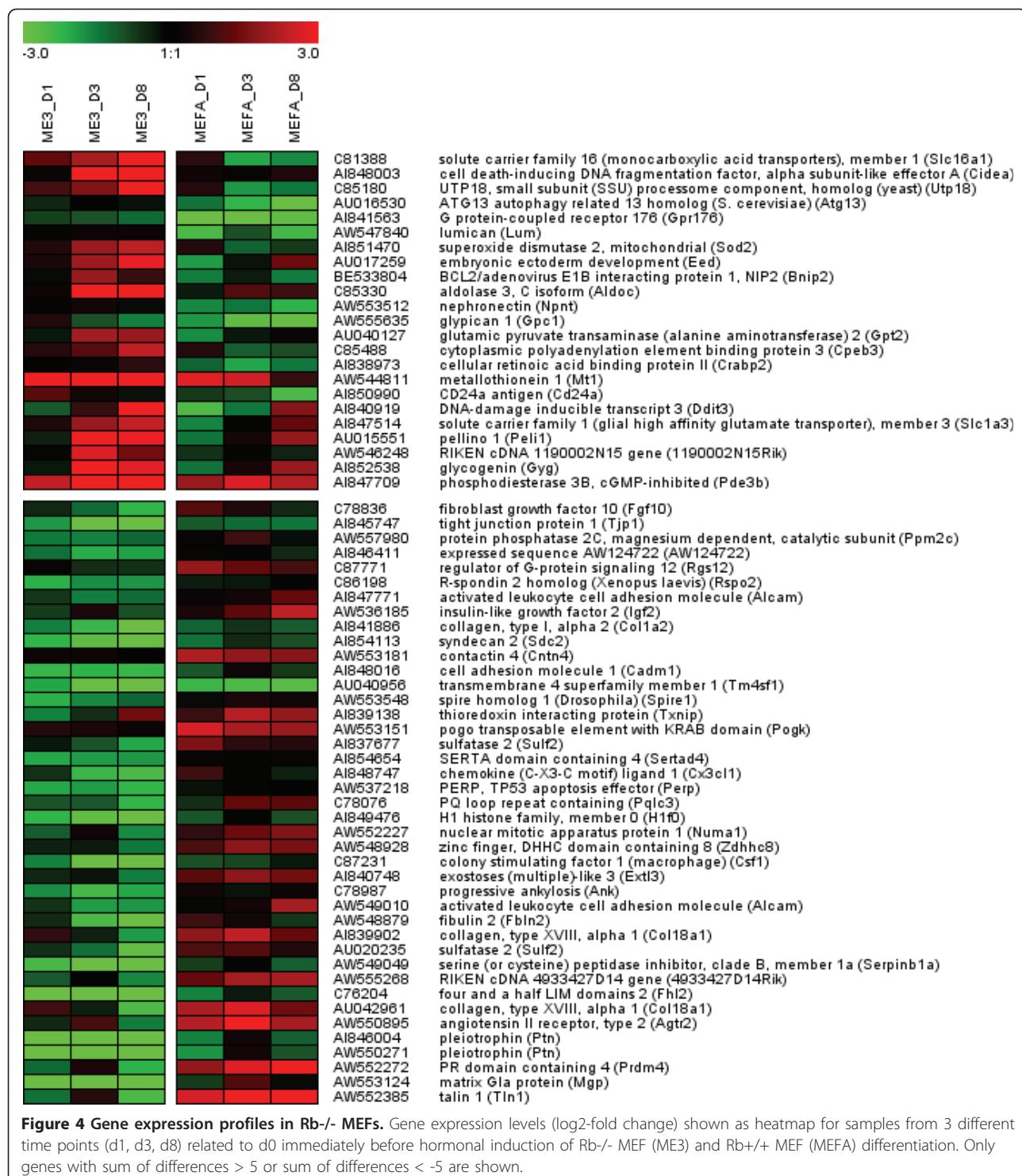
In this work we studied adipocyte differentiation of wild-type MEFs and Rb-/ MEFs into mature adipocytes using microarrays and comprehensive bioinformatics analyses. We identified novel candidates involved in early adipogenesis as well as high-confidence genes targeted by Pparg. Our findings corroborate the brown-like phenotype of the Rb-/ MEFs. Based on the analyses we also propose the existence of a feed-back loop between Pparg and Foxo1 adding a novel facet to the regulatory network orchestrating adipocyte differentiation.

The data and the analysis provide a starting point for further experimental studies to identify target genes for pharmacological intervention and ultimately remodeling of white adipose tissue into brown adipose tissue.

Materials and methods

Adipocyte differentiation of wild-type (Rb++) and Rb-/ mouse embryonic fibroblasts

Generation of Rb-/ MEFs and wild-type MEFs were previously performed at the Division of Cancer Biology, Danish Cancer Society, Copenhagen, DK and described in [49]; detailed information on the origin of these cells and disruption of the Rb-1 gene in mice can be found in [50]. A batch of the Rb-/ and Rb++ MEFs were



kindly provided by Jiri Lukas as already used in other studies (e.g. [14]). No animals or human subjects were directly involved in the current study that would make ethical approval and consent required. Wild-type mouse embryo fibroblasts (MEFs) were differentiated at the Department of Biochemistry and Molecular Biology,

University of Southern Denmark, Odense, DK according to the protocol described in [13]. Briefly, cells were induced to differentiation by the MDI protocol (1 μM dexamethasone (DEX), 0.5 mM isobutyl methylxanthine (MIX), and 5 μg/ml insulin). DEX and MIX were omitted after day 2, but insulin was added throughout

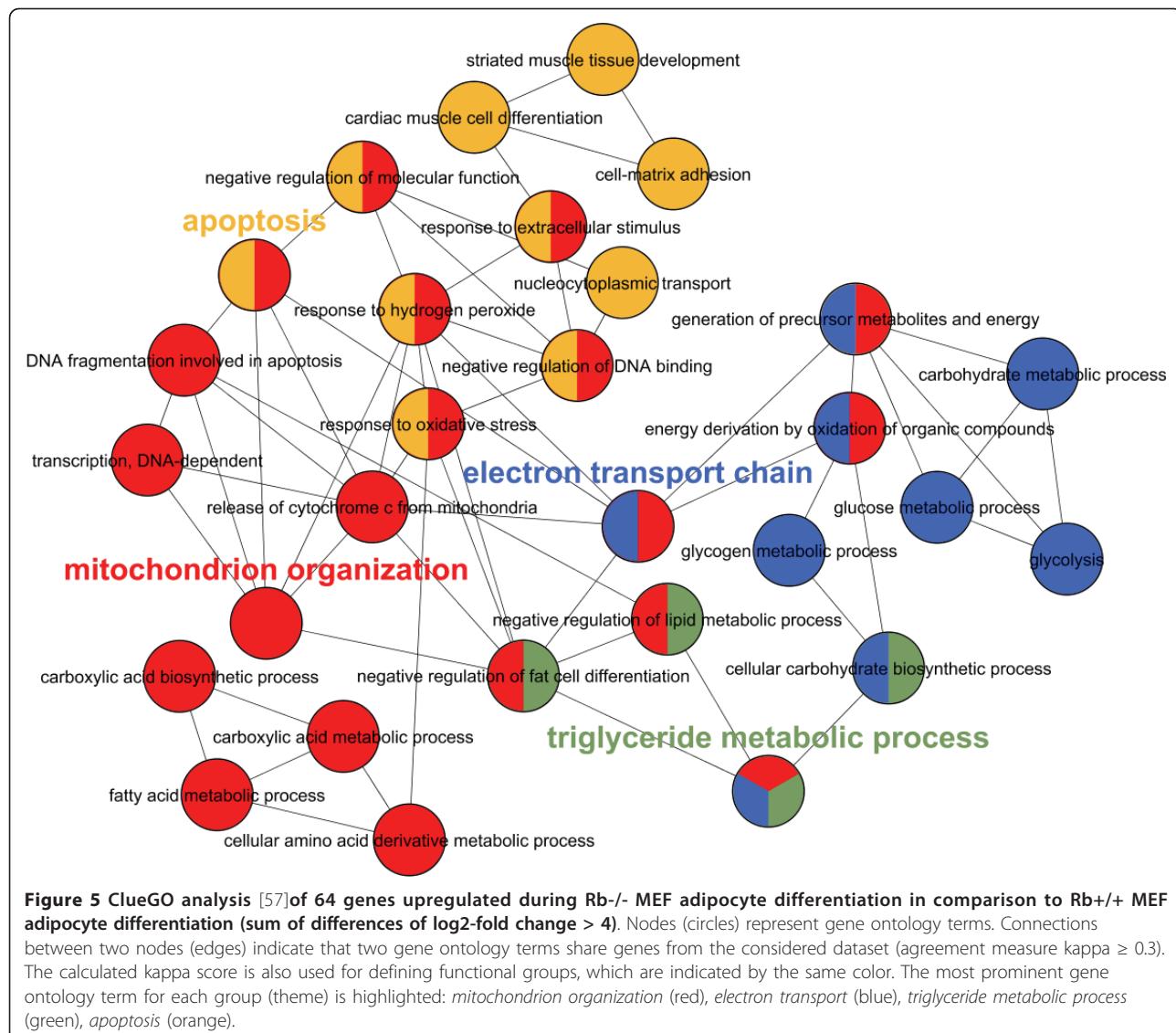


Figure 5 ClueGO analysis [57] of 64 genes upregulated during Rb-/- MEF adipocyte differentiation in comparison to Rb+/+ MEF adipocyte differentiation (sum of differences of log₂-fold change > 4). Nodes (circles) represent gene ontology terms. Connections between two nodes (edges) indicate that two gene ontology terms share genes from the considered dataset (agreement measure kappa ≥ 0.3). The calculated kappa score is also used for defining functional groups, which are indicated by the same color. The most prominent gene ontology term for each group (theme) is highlighted: *mitochondrion organization* (red), *electron transport* (blue), *triglyceride metabolic process* (green), *apoptosis* (orange).

the differentiation. Cells were passaged and differentiated in AmnioMax (Gibco) supplemented with 7,5% fetal bovine serum. Nutrition media were changed every second day. Rb-/- MEFs and wild-type MEFs (Rb+/+) for comparison were differentiated using the identical protocol, but adding Pparg ligand BRL49653 (0.5 μM) to the induction cocktail throughout the whole differentiation process. All differentiation experiments were performed in triplicate. Cells were harvested and total RNA was isolated at the preconfluent stage and at seven time points after induction (0, 1, 2, 3, 4, 6, and 10 days) with TRIzol reagent (Invitrogen). For the Rb-/- and Rb+/+ MEF differentiation, cells were harvested immediately before (reference) and at three time points (1, 3, and 8 days) after induction. For each independent experiment RNA was pooled from three different culture dishes for each time point and from twenty dishes for each

reference. The quality of the RNA was checked using Agilent 2100 Bioanalyzer RNA assays (Agilent Technologies) by inspection of the 28S and 18S ribosomal RNA intensity peaks.

Gene expression and data analysis

The mouse cDNA microarrays and the hybridization protocols used have been described previously [2]. Briefly, the spotted microarrays contain 27,000 elements with mouse cDNA clones representing 16,000 different genes (UniGene clusters). 10-25 micrograms (dependent on replicated experiment) of total RNA from each time point was reverse transcribed into cDNA, which was then indirectly labeled with Cy5 and Cy3, respectively. To account for technical variation, procedures were repeated using the same samples with reversed dye assignment. The microarrays were pre-hybridized with

5× SSC, 0.1% SDS, 1% BSA. Pair-wise labeled cDNA samples were combined and 20 μg of mouse Cot1 DNA and 20 μg of poly(A) DNA were added. The mixture was hybridized onto the slides overnight at 42°C. Following washing, slides were scanned with a GenePix 4000B microarray scanner (Axon Instruments) at 10 μm resolution. The resulting TIFF images were analyzed with GenePix Pro 4.1 software (Axon Instruments). Features were filtered for low-quality spots. Following subtraction of the local background, the arrays were global median and dye-swap normalized using ArrayNorm [51] and the resulting ratios log₂ transformed. All experimental parameters, raw data, and transformed data were uploaded to the microarray database MARS [52] and submitted via MAGE-ML export to a public repository (ArrayExpress [53], Accession Nos. E-MARS-12/E-MARS-13 and A-MARS-7). After outliers were removed the median expression values of replicated ESTs on the microarray were calculated. Only ESTs with $z > 1.5$ and significant replication ($p < 0.05$) in at least one time point using the z test and implying standard normal distribution were considered differentially expressed and used for further analysis [54,55]. Expression values and z scores for differentially expressed ESTs were averaged over biological replicates and annotated using the RefSeq database (MegaBLAST E-value < 1E-30) and Entrez Gene database. Expression profiles for MEF data and 3T3-L1 data [2] were combined based on the same EST ID and profiles were grouped into 6 clusters by k-means clustering and validated by principal component analysis. Gene ontology term overrepresentation analysis for the different gene sets (clusters) were performed using Fisher exact test and p-values were adjusted for multiple hypothesis tests based on the Benjamini-Hochberg method (<http://genome.tugraz.at/ORA>). Several distance measures between the expression profiles of the Rb-/ MEF (ME3) and Rb+/+ MEF (MEFA) differentiation models (sum of differences, Manhattan distance, Euclidean distance, Pearson correlation coefficient) were calculated. Over-represented gene ontology (GO) terms for up- (sum of differences > 4) or downregulated (sum of differences < -4) genes were identified using DAVID [56] based on gene symbols. Up-regulated genes were functionally grouped into gene ontology networks using the Cytoscape plug-in ClueGO [57]. All calculations were implemented in PERL 5.8.0 or R 2.10.1, cluster analyses and visualizations were performed using Genesis [58]. Gene set enrichment analysis (GSEA) [59] was performed using a tool from <http://www.broadinstitute.org/gsea>.

qPCR analysis

Microarray expression results were confirmed with qPCR. cDNA was synthesized from 1.25 μg total RNA

in 20 μl using random hexamers and SuperScript II reverse transcriptase (Invitrogen). The design of LUX™ primers was done using the Invitrogen web service. Quantitative RT-PCR analyses were performed starting with 12 ng reverse transcribed total RNA, with 0.5× Platinum Quantitative PCR SuperMix-UDG (Invitrogen), with a ROX reference dye, and with a 200 nM concentration of both LUX™ labeled sense and antisense primers (Invitrogen) in a 25 μl reaction. Ribosomal 18S RNA amplifications were used to account for variability in the initial quantities of cDNA. The relative quantification with respect to the calibrator (preconfluent stage) was determined using the ΔΔC_t method and compared with the normalized ratios resulting from microarray experiments. To better characterize Rb-/ MEF (ME3) and Rb+/+ MEF (MEFA) cells, triplicate-qPCR analyses were performed with 4.5 ng cDNA, 200 nM primer, and Platinum SYBR green mastermix (Invitrogen). Data were analyzed using a real-time PCR management and analysis system [60]. Values were normalized to Uxt, Ct values and PCR efficiencies were derived as in [61]. All qPCR analyses were performed on an ABI PRISM 7000 detection system (Applied Biosystems). Primer sequences can be found in Additional file 2.

In silico analysis of transcription factor binding sites

Statistically overrepresented ($p < 0.05$) potential transcription factor binding sites (TFBS) were identified by *in silico* analysis of genomic regions from -3 kb to + 2 kb around transcription start sites of genes within each cluster and all RefSeq transcripts using known (TRANSFAC [62], JASPAR [63]), and newly compiled position frequency (weight) matrices (<http://genome.tugraz.at/Logo>). Potential binding sites were identified if the similarity score for each PWM was equal or above the threshold (allowing 1 binding site with 10kb of mouse coding sequences) based on an implementation of the MatInspector algorithm [64]. The number of targets (showing at least one TFBS) within the gene set (cluster) was tested against the number of targets from all RefSeq transcripts using Fishers exact test and p-value was adjusted for multiple hypothesis testing based on the false discovery rate according to the method of Benjamini-Hochberg [65]. Genome organization around genes (genomic region from 20 kb upstream of transcription start to 20 kb downstream of transcription end) showing similar expression profile as Pparg were constructed using UCSC genome browser (NCBI37/mm9), custom track data from chromatin immune precipitation followed by sequencing at day 6 [42] or microarray analysis at day 10 [43] during 3T3-L1 adipocyte differentiation, and Pparg-Rxra direct repeats 1 (DR1) motifs (similarity score > 0.90) (see additional website <http://icbi.at/adipo>).

Additional material

Additional file 1: Gene expression levels (log₂-fold change) of 1579 in 6 cluster grouped ESTs at several time points during adipocyte differentiation of MEFs and 3T3-L1 cells (as visualized in Figure 3).

ESTs were annotated according to the mouse RefSeq database (NCBI MegaBLAST E-value < 1E-30) and Entrez Gene database. Unique genes within each cluster were identified and overrepresented gene ontology (GO) terms are provided.

Additional file 2: Results and primer sequences from q-RT-PCR analysis for validation of microarray experiments and to identify expression levels for specific genes during ME3 and MEFA differentiation.

Additional file 3: Statistically overrepresented ($p < 0.05$) potential transcription factor binding sites (TFBS) identified by *in silico* analysis of genomic regions from -3 kb to + 2 kb around transcription start sites of genes with increasing gene expression at late stages of adipocyte differentiation (cluster F) using position frequency (weight) matrices from TRANSFAC and JASPAR.

Corresponding sequence logos from <http://genome.tugraz.at/Logo> are shown and number of targets within the dataset (cluster) is given and tested against the number of targets from all RefSeq transcripts using Fishers exact test. Note only transcription factor motifs are considered which are exclusively overrepresented in cluster F.

Additional file 4: Gene expression levels (log₂-fold change) of 3118 ESTs for samples from 3 different time points (d1, d3, d8) related to d0 immediately before hormonal induction of Rb-/ MEF (ME3) and Rb+/+ MEF (MEFA) differentiation.

Several distance measures between the expression profiles of the two cell differentiation models (sum of differences, Manhattan distance, Euclidean distance, Pearson correlation coefficient) are provided. Unique genes upregulated (sum of differences > 4) and downregulated (sum of differences < -4) in the ME3 cell differentiation model compared to the MEFA differentiation are emphasized and overrepresented gene ontology (GO) terms are provided.

Additional file 5: Gene set enrichment analysis of the gene list from comparison of Rb-/ MEFs and Rb+/+ MEFs during adipocyte differentiation ranked by sum of differences in a gene set of significantly upregulated (log₂-fold change > 2) genes in brown adipose tissue versus white adipose tissue [46] and in a gene set with significantly upregulated (log₂-fold change > 1) genes in brown versus white preadipocytes at the differentiating stages [47].

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Authors' contributions

Performed and conceived MEF and Rb-/ MEF cell culture experiment: CJ, JBH, LM, KK. Oversee qPCR analysis: AMK, HH. Generation of microarray

slides: AP. Microarray experiments: HH. Bioinformatics and statistical analysis: MF, HH. Data analysis and interpretation: RHW, AMK, AP, JBS, SRB, ZT, HH. Wrote the paper: RHW, ZT, HH. All authors read and approved the manuscript.

Competing interests

The authors declare that they have no competing interests.

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