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Identification of the Rock-dependent transcriptome in rodent fibroblasts

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Abstract

Rock proteins are Rho GTPase-dependent serine/threonine kinases with crucial roles in F-actin dynamics and cell transformation. By analogy with other protein kinase families, it can be assumed that Rock proteins act, at least in part, through the regulation of gene expression events. However, with the exception of some singular transcriptional targets recently identified, the actual impact of these kinases on the overall cell transcriptome remains unknown. To address this issue, we have used a microarray approach to compare the transcriptomes of exponentially growing NIH3T3 cells that had been untreated or treated with Y27632, a well known specific inhibitor for Rock kinase activity. We show here that the Rock pathway promotes a weak impact on the fibroblast transcriptome, since its inhibition only results in changes in the expression of 2.3% of all the genes surveyed in the microarrays. Most Y27632-dependent genes are downregulated at moderate levels, indicating that the Rock pathway predominantly induces the upregulation of transcriptionally active genes. Although functionally diverse, a common functional leitmotiv of Y27632-dependent genes is the implication of their protein products in cytoskeletal-dependent processes. Taken together, these results indicate that Rock proteins can modify cytoskeletal dynamics by acting at post-transcriptional and transcriptional levels. In addition, they suggest that the main target of these serine/threonine kinases is the phosphoproteome and not the transcriptome.

Keywords

Rock; Rho/Rac GTPases; Microarray; Gene expression; Transcription; Cytoskeleton

Introduction

The Rho/Rac family is a large group of GTP-binding proteins specialised in the regulation of a wide spectrum of cellular functions such as cytoskeletal organisation, cell proliferation, vesicle trafficking and cytokinesis [1–6]. Rho/Rac proteins are regulated by extracellular stimulus-dependent changes in their bound guanosine nucleotides. In non-stimulated cells, these proteins are bound to GDP molecules and in an inactive conformation [1,7]. In addition, they remain sequestered in the cytosol due to their interaction with Rho GDP dissociation inhibitors (RhoGDIs) [1,8,9]. In stimulated cells, these proteins are released from RhoGDIs, translocate from the cytosol to cellular membranes and undergo the exchange of GDP by GTP. This exchange of nucleotides promotes a conformational change in the GTPase switch regions that, in turn, allows the binding of downstream effectors [1]. This activation step, as well as

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its subsequent inactivation by hydrolysis of the bound GTP molecules, is catalysed by guanosine nucleotide exchange factors and GTPase activating proteins, respectively [10,11]. According to structural criteria, Rho/Rac proteins can be subdivided in Rho (RhoA, RhoB, RhoC), Rac (Rac1, Rac2, Rac3, RhoG) and Cdc42 (Cdc42, TTF) subfamilies [1]. Within the Rho subfamily, RhoA is perhaps the best characterised in terms of three-dimensional structure, effectors and its participation in normal and pathological-related biological responses [1,12– 15].

The elucidation of the regulation and function of RhoA effectors is important to understanding the intracellular pathways that control RhoA-dependent cellular, physiological and pathological responses. Two of the main RhoA effectors are the highly related serine/threonine kinases RockI (also known as Rok β and $p160^{Rock}$) and RockII (also referred to as both Roka and Rho kinase) [16,17]. These proteins become activated during signal transduction by the binding to Rho subfamily members [16,17] and/or second messengers such as arachidonic acid [18] and sphingo-sylphosphorylcholine [19]. Their activities are also subjected to negative regulation by specific subsets of GTPases (RhoE, Gem, Rad) [20,21] and cell cycle inhibitors $(p21^{Cip1})$ [22].

Rock proteins induce intracellular pathways that mediate the formation of stress fibres and focal adhesions, thereby participating in cell-to-cell and cell-to-substratum adhesion, cell migration and motility, phagocytosis and neurite retraction [16,17]. They also work in cell division and cytokinesis processes by regulating centrosomal functions and actomyosin ring contraction, respectively [16,17]. Deregulated signalling outputs from these kinases appear to be important for some pathologies, such as hypertension, Alzheimer's disease and cancer [12,14–17]. Demonstrating the role of these proteins in these diseases, it has been shown that the use of chemical inhibitors for Rock proteins alleviates cardiovascular pathologies such as pulmonary hypertension, vasospasms and angina pectoris [14,16,17]. Rock inhibitors also block tumorigenesis *in vitro* [23] and appear to be potentially useful for the treatment of other medical conditions including Alzheimer's disease, stroke and neuropathic pain [17].

Several Rock downstream targets have been identified, including regulators of the F-actin cytoskeleton (myosin light chain (MLC), the MLC phosphatase, Lim kinases 1 and 2), intermediate filament components (vimentin, glial fibrillary acidic protein and neurofilaments) and microtubule-associated proteins (Tau, microtubule-associated protein 2) [16,17]. Whereas the phosphorylation of MLC and its phosphatase by Rock proteins promotes the formation of F-actin fibres, the phosphorylation of other protein classes appears to induce neurofilament disassembly and to halt microtubule polymerisation. Thus, the phosphoproteome induced by Rock proteins is fully consistent with the assigned roles of these proteins in cell migration and morphology [16,17].

Similar to other serine/threonine kinases involved in signal transduction (i.e., Erk, p38MAPK), it is possible that Rock could also promote the long-term regulation of gene expression. Consistent with this view, it has been shown that Rock activity is important for the stimulation of c-Myc by the constitutively active, oncogenic version of RhoA (Q63L mutant) [24,25] and for the expression of a small subset of the transcriptome of NIH3T3 cells transformed by the chronic expression of the *rhoA* oncogene [24]. Other studies have also shown that the expression of specific Rho A^{Q63L} -dependent genes is abrogated upon inhibition of the Rock pathway [26,27]. In the present study, we aimed at expanding these results to non-transformed fibroblasts. To this end, we used microarray technology to assess the effect of Y27632, a chemical inhibitor commonly used to block Rock kinase activity [28], in the transcriptome of exponentially growing NIH3T3 cells. This cell line has been widely utilised before for the characterisation of the biological properties of both Rho and Rock proteins. Previous observations by us and others have shown that Y27632 treatments inhibit several Rock-

dependent responses in this cell line, including MLC phosphorylation and stress fibre formation [23,24]. We report here the results obtained from this research avenue.

Materials and methods

Cell lines

Murine NIH3T3 cells were grown under standard temperature/ $CO₂$ conditions in Dulbecco's modified Eagle's medium supplemented with 1% L-glutamine, 1% penicillin/streptomycin and 10% calf serum. All tissue culture reagents were obtained from Invitrogen. When appropriate, cells were treated for 24 h with $10 \mu M$ Y27632 (Cal-biochem) to inhibit endogenous Rock proteins. RhoA-transformed cells have been described before [24]. To confirm the effectiveness of Rock inactivation in this experimental setting, parallel cultures of NIH3T3 and RhoA-transformed cells were analysed by immunoblot and immunofluorescence techniques to corroborate the expected inhibition of the phosphorylation of the myosin light chain and the disassembly of stress F-actin fibres in Y27632-treated cells, as indicated and shown before [24].

Microarray experiments and data analysis

Microarray analyses were performed using RNAs obtained from seven and five independent experiments of untreated and Y27632-treated NIH3T3 cells, respectively. In each independent experiment, three 10-cm diameter plates containing exponentially growing cultures were used to generate the total RNA used in the microarray studies. To this end, cultured cells were washed with phosphate-buffered saline solution and total cellular RNAs isolated using the RNeasy kit (Qiagen) according to the supplier's specifications. The quantity and quality of the total RNAs obtained were determined using 6000 Nano Chips (Agilent Technologies). Total RNA samples (4 µg) were then processed for hybridisation on MGU75Av2 microarrays (Affymetrix) using standard Affymetrix protocols at the CIC Genomics and Proteomics Facility. Normalisation, filtering and analysis of the raw data obtained from microarrays were carried out with the Bioconductor software (www.bioconductor.com) using the *ReadAffy* package and the RMA application. We considered a gene to be differentially expressed when exhibiting a signal ≥ 100 and its fold change respect to the levels of expression of untreated NIH3T3 cells was $\geq \pm 1.5$ and with *p* values ≤ 0.01 . Statistical analyses were performed using F-statistics.

For the graphical presentation of microarray data, we performed hierarchical clustering analysis using the WP-GA average-linkage and the standard correlation similarity metric method with the J-Express application. Functional annotation of gene functions was performed manually using internet-available databases such as those maintained by the NCBI [\(www.ncbi.nlm.nih.gov/sites/entrez?db=omim\)](http://www.ncbi.nlm.nih.gov/sites/entrez?db=omim) and the Weizman Institute of Science (Rehovot, Israel; [www.genecards.org\)](http://www.genecards.org). The identification of interactive networks of proteins and common functions was done using the Ingenuity Pathways Analysis program, a webdelivered application that enables discovery, visualisation and exploration of biological interaction networks and biological processes [\(www.ingenuity.com\)](http://www.ingenuity.com) [29]. In this case, we considered a network as significant when it fulfilled the following criteria: (i) a minimal score of 15; (ii) at least 10 proteins participating in direct, functional interactions inside the network. In addition, and to reinforce the strength of the functional relationships, we only took into consideration direct, not indirect, relationships among the network components. Bioinformatically identified networks were edited manually to sieve out proteins that, according to published data, did not have a coherent or well defined functional relationship with the other network components.

The comparison of transcriptomes between Y27632-treated non-transformed and RhoA^{Q63L}transformed NIH3T3 cells was done using the previously published data on the $RhoA^{Q63L}$ dependent transcriptome [24].

Real-time quantitative RT-PCR

Exponentially growing cells treated and non-treated with Y27632 were lysed and total RNA extracted using the RNAeasy kit (Qiagen). RNAs were quantified by loading aliquots into 6000 Nano Chips. Quantitative polynucleotide chain reactions were performed using the Quanti-Tect SYBR Green RT-PCR kit (Qiagen). 18S rRNA primers were used as controls for both loading and quantitation of relative expression levels of the genes tested. Amplifications were performed using the iCycler machine (Bio-Rad). Raw data were analysed using the iCycler iQ Optical System software (version 3.0a, Bio-Rad). In other cases, quantitative RT-PCR experiments were conducted using a microfluidic card (Applied Biosystems) service available at the Program for Genomics, Proteomics and Bioinformatics of the Spanish Network of Cancer Groups.

Results and discussion

We made use of Affymetrix microarray technology to identify the transcriptomal changes induced by culturing exponentially growing NIH3T3 cells with the Rock inhibitor Y27632 for 24 h. The reason for selecting this time point was two-fold. On the one hand, it ensured effective Rock inhibition, since the dephosphorylation of MLC is detected already 6 h after addition of Y27632 and remains at low levels thereafter [24]. On the other hand, this early time point allowed us to select for primary Rock transcriptional targets rather than detecting secondary transcriptomal changes derived from the deregulation of the expression of putative Rockdependent transcriptional factors. We also cultured cells in the presence of serum and under non-confluent conditions (approx. 70% confluency) to avoid the activation of strong genetic programs related to serum withdrawal or contact inhibition that may obscure the detection of the Rock-dependent transcriptome [30]. In addition, we isolated total RNAs from seven (in the case of NIH3T3 cells) and five (in the case of Y27632-treated cells) independent cell cultures in order to make a robust statistical treatment of the data generated possible. Total RNAs samples obtained under those conditions were converted into biotinylated-cRNA probes and hybridised independently to Affymetrix Genechip MG U74Av2 arrays, thus allowing the monitoring of the expression status of \approx 12,500 mouse genes.

The results from these microarray experiments indicated that the chemical inhibition of Rock led to changes in approximately 2.3% (289 genes) of all genes probed in the arrays (Fig. 1A, B and Table 1). Interestingly, these changes involved mostly the downregulation of transcriptionally active genes (76.1% of all responsive genes), indicating that Rock activity is required primarily to maintain transcription from specific gene subsets in mouse fibroblasts. Instead, the inhibition of Rock determined the activation of a much smaller group of 69 genes. To confirm the microarray results, the expression of seven of the upregulated and ten of the downregulated genes was re-evaluated using real-time quantitative RT-PCR analysis. These analyses confirmed that these 17 genes were indeed Rock-dependent (Fig. 2). Interestingly, we also detected genes identified as Rock targets such as *cyr61, c-myc* and *cyclin D1* [24,26, 27], further supporting the validity of our microarray data.

We observed that the Rock-dependent genes were distributed following a Poisson-like distribution when classified in terms of overall fold-change variations (Fig. 1C). Thus, the majority of up- and downregulated genes displayed modest changes (1.5–2.1-fold) when their expression levels were compared between Y27632-treated and untreated NIH3T3 cells. Instead, only a small minority of genes showed fold-change variations outside that interval (Fig. 1C). Upregulated and downregulated genes followed similar trends, although we only

observed upregulated loci in the subset of genes displaying variations larger than 6-fold (Fig. 1C). We observed that the genes undergoing the largest upregulation encoded either chemokines (Ccl13, Cxcl6, Ccl7) or secreted factors (Cp and the component 3 of the complement) (Table 1). Instead, the genes with largest repressions encoded for cytoskeletalrelated proteins such as actin α 2, Ptx3, Thbs1 and Ogn (Table 1).

To establish an overview of the transcriptional changes induced by Rock inhibition, we assigned the 289 identified genes regulated by Y27632 to 19 different functional groups using manual annotation procedures (Fig. 3A, Table 1). This analysis revealed that the main functional groups targeted by the Y27632 include those corresponding to extra-cellular matrix proteins, membrane-localised proteins, cytoskeleton, transcriptional regulation and metabolism. With the exception of the apoptosis and cell cycle-related class, all the other functional groups contain a larger number of downregulated than of upregulated genes. In fact, five of those classes (growth factors, protein biosynthesis, heat shock, metabolism, cytoplasmic/nuclear transport) contain only repressed genes (Fig. 3A, Table 1). Interestingly, the largest percentage of upregulated genes is seen in the immune-related (83%), cell cycle (63%), ligand (58%) and membrane receptor (83%) subclasses (Table 1).

The main problem with functional annotation is that it groups genes according to functional similarity or relatedness. Due to this, this type of analysis usually oversees other important biological information, such as the interconnectivities established among genes of different functional classes, homogeneous alterations of signal transduction pathways, etc. To surmount this problem, we subjected our microarray data to further bioinformatic characterisation using the Ingenuity program. This web-based software allows the identification of common biological processes and molecular networks because it relates each gene entry with a comprehensive database of known protein–protein, transcriptional or enzymatic relationships available for = 10,000 mammalian proteins [29]. At the level of molecular networks, the analysis of the Y27632-targeted transcriptome using this bioinformatics package indicated changes in the expression of genes whose protein products are involved in integrin and c-Myc function (Fig. 3C,D). The composition of the first network is, however, limited to specific integrin subunits and proximal cytoskeletal components (see below), indicating that the signalling cascades located down-stream of integrins are not touched by the inhibition of the Rock route. The detection of the c-Myc network is of interest, because we have shown before that the overexpression of $RhoA^{Q63L}$ promotes this network whereas the inhibition of Rock downmodulates it in *rhoA*-transformed cells [24]. A third molecular network contained a larger number of protein constituents (30 in total) (Fig. 3B). This network can be further subdivided into two main branches, one that is loosely related with the Ccl7 and Ccl13 chemokines and one of its repressors (the transcriptional factor Bcl6) and another branch connecting the downmodulation of specific nuclear proteins (Xbp1, Nr2f2) with the repression of the expression of extracellular proteins such as collagen, serpin and a lipoprotein lipase (Lpl). Unlike the other two networks, this third molecular conglomerate does include nuclear, cytosolic, membrane and extracellular-located molecules, suggesting that its targeting by the Rock pathway may have some signalling purpose. To our knowledge, however, this network has not been linked to any established biological process so far. Quantitative RT-PCR experiments demonstrated that the selected elements of these three networks are indeed deregulated upon treatment of NIH3T3 cells with the Rock inhibitor (Fig. 4). At the level of biological processes affected, the bioinformatic analysis indicated that the inactivation of Rock by Y27632 alters, in a statistically significant manner (*p*≤0.001), genes whose protein products are primarily in charge of cellular functions usually regulated by these serine/threonine kinases such as cellular movement (migration, chemotaxis, transmigration, haptotaxis, invasion, scattering), cell morphology and the extracellular matrix. The products of these genes were also either directly (i.e., c-Myc) or indirectly (rest of genes) linked to cell growth and survival processes. However, this latter functional category probably has poor significance from a

biological point of view, because we have not detected a large pool of genes directly activated or repressed in proliferating cells (i.e., proteins directly involved in the cell cycle machinery, replication origins, DNA synthesis, etc.) (Table 1). This is consistent with previous observations by us and others indicating that Y27632 treatments do not significantly alter the proliferation rates of fibroblasts [23,24]. Based on these results, the only obvious common feature observed among these transcriptomal changes is their relationship, direct or indirect, with integral and regulatory components of cellular components usually regulated by Rock proteins such as F-actin cytoskeleton, microtubules and cell movement-related processes. Consistent with this view, it was observed that cytoskeleton-related proteins (actinin, vinculin, calponin, talin, spectrins, actin itself), cytoskeletal regulators (thy-mosin β4, myosin subunits, integrins, transgelin, catenin, Nap125, dynamin, caveolin) and microtubule-related molecules (Macf1, kinesins) showed up in most of the networks and pathways picked up by the Ingenuity software.

We have previously shown that the inhibition of Rock in RhoA^{Q63L}-transformed NIH3T3 cells with Y27632 also provokes minor changes in the cell transcriptome of these oncogenically transformed cells [24]. The similarity in the experimental conditions used in that work and in the current study made it possible to compare side by side the effects of Y27632 in transformed and non-transformed cells. This analysis indicated that Y27632 induced larger expression changes in the transcriptome of non-transformed (298 genes) than in RhoA^{Q63L}-transformed (179 genes) cells. Moreover, we have observed that the gene subsets targeted by the Rock inhibitor in the parental and the Rho A^{Q63L} -transformed NIH3T3 cells were significantly different, since these two transcriptomes only show 70 coincident target genes. These shared genes showed the same change pattern and belonged to both the upregulated (17 genes) and downregulated (53 genes) classes. This subset of genes included the c-Myc network detected in the Ingenuity analysis, although this interactive molecular network has a larger number of components in RhoA-transformed cells than in the non-transformed parental cells [this work, 24]. Four additional genes, although targeted by Y27632 in both cell types, showed opposite change patterns in normal and $RhoA^{Q63L}$ -transformed cells. The functional classes deregulated by the Y27632 treatment that displayed more disparity between non-transformed and transformed cells were those related with membrane-located, cell adhesion-related proteins, cell cycle, DNA replication and electron transporter. The classes showing more coincident expression changes encompassed those related to extracellular, cell-adhesion-related functions, extracellular ligands, cytoskeleton, regulation of transcription and proteins with unassigned functions. These data indicate that the impact of the Rock pathway on the transcriptome is always small regardless of whether fibroblasts have normal or exacerbated RhoA activity levels. However, the type of genes targeted by this signalling route is significantly different depending on the transformed status of these cells.

In summary, these results indicate that the Rock pathway has a rather weak impact in the transcriptome of normal fibroblasts. Therefore, they are consistent with the idea that the main signalling purpose of this route is to induce phosphoproteomal rather than transcriptomal changes in the cell. Furthermore, the relative small fold change variations found in the majority of Y27632-targeted genes indicated that the transcriptional action of the Rock pathway relies mainly on modulating the levels of activity of already active genes rather than on turning on previously silent loci or turning off active genes. Interestingly, most Rock-dependent transcriptional targets are downmodulated upon inhibition of Rock activity, indicating that the Rock pathway is oriented fundamentally to the upregulation of genes under exponentially growing conditions. Finally, we have observed that the main transcriptional targets affected by the blockage of Rock activity in fibroblasts are related with processes of cell movement, cell shape and F-actin dynamics. Thus, the post-transcriptional action of Rock proteins on the F-actin cytoskeleton appears to be coordinated, in the long-term, with the modulation at the transcriptional level of genes involved in the regulation of those components of the cell

architecture. It will be important in the future to complement these studies with others focused on the cell proteome and phosphoproteome to get a comprehensive view of the effects and impact of Rock function in the biology of the mammalian cells.

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Fig. 1.

Transcriptomal changes induced by Y27632 in exponentially growing NIH3T3 cells. **A** Hierarchical cluster diagram of the 289 genes whose expression levels changed in Y27632 treated cells. Each column represents one experiment and each row a gene. Varying levels of expression are represented on a scale from dark blue (lowest expression) to dark red (highest expression). Note that expression values are represented as signal log ratio numbers (SLR) and that, therefore, the total fold change value is obtained from 2SLR. The experiment number is shown at the top of each column. **B** Gene graphs showing the induced (red) and repressed (blue) genes in Y27632-treated cells. In each category, the expression values of all deregulated genes are represented as SLR (considering that fold change is 2SLR; *y*-axis) obtained in each experimental sample (*x*-axis). The total number of upregulated and downregulated genes in each category is indicated on the right of each panel. **C** Histogram showing the number of up-

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(red) and downregulated (blue) genes with a given expression fold change value in Y27632 treated cells

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Fig. 2.

Corroboration of Affymetrix data by quantitative RT-PCR. **A, B** Expression levels of the indicated up-regulated (A) and downregulated (B) mRNAs by the Y27632 treatment were determined by either microarray (A, grey bars) or quantitative RT-PCR (Q, black bars). Values are expressed as fold change of the appropriate gene with respect to the transcript levels found in untreated NIH3T3 cells

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Fig. 3.

Functional annotation and characterisation of Y27632-dependent genes. **A** Classification of up- (red) and downregulated (blue) genes by the Y27632 treatment according to general biological functions. **B**–**D** The molecular networks identified using the Ingenuity database in the Y27632-affected transcriptome. Nodes are colour-coded in red (upregulated) or green (downregulated) according to their fold change values

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Fig. 4.

Corroboration of the molecular networks identified in Fig. 3 by quantitative RT-PCR. **A**–**C** RT-PCR-determined expression levels of selected genes belonging to the molecular networks shown in Fig. 3B (A), 3C (B) and 3D (C). Values are expressed as fold change of the appropriate gene with respect to the transcript levels found in untreated NIH3T3 cells

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a

Table 1 List of genes whose expression is upregulated or downregulated by Y27632 in NIH3T3 cells

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Cell division cycle 25 homologue C

Cdc25c

12532

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 $\begin{array}{c} 0.0013 \\ 0.0002 \end{array}$

 $0.62\,$ 0.63

11966 Atp \sim 121.8 of transporting, isoform 2 \sim 134 subunit B, isoform 2 \sim 0.12 \sim 0.13 \sim 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 0.0013 66335 Atp6v1c1 Atpase, H+ transporting, V1 subunit C, isoform 1 0.63 0.0002

Atpase, H+ transporting, V1 subunit B, isoform 2 Atpase, H+ transporting, V1 subunit C, isoform 1

Atp6v1b2 Atp6v1c1

11966 66335

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 $<\!\!1.00\mathrm{E}\textrm{-}04$

 1.64 0.51

Serine (or cysteine) proteinase inhibitor, clade A-3G

Ribonuclease, mase A family 4

0.0043

 $\begin{array}{c} 0.0001 \\ 0.0001 \\ 0.0003 \\ 0.0012 \end{array}$

 $\begin{array}{c} 0.6 \\ 0.6 \\ 0.59 \end{array}$

16203 No.59 $\frac{1}{2}$ New Assumption and $\frac{1}{2}$ amidation $\frac{1}{2}$ or $\frac{$ 260000 0.00000 0 6000000 RAN binding protein 9 0.000 μ RAN binding protein 9 0.000 μ RAN binding protein 9 0.00 μ 2112060 0.59 Rcn2 Reticulois 2 R eticulois 2 R eticulois 2 R eticulois 2 R α_{31} 9714 α_{10} 19714 α_{11} 1974 α_{12} anset A family 4 α_{13} anset A family 4 20715 Serpina3g Serine (or cysteine) proteinase inhibitor, clade A-3 g 1.64 and 20 mins of the A-3G 1.64 and 20

N-terminal Asn amidase

RAN binding protein 9 PHD finger protein $17\,$

Reticulocalbin $2\,$

Ntan1
Phf17
Ranbp9
Ren2
Rnase4
Rnase4

269424 18203

56705

26611
319714
20715

 0.59

Genes have been classified into 19 different functional groups. The locus identification number (Locus ID), the gene symbol (gene) and spelled out designation of each gene are shown. For the sake of ^aGenes have been classified into 19 different functional groups. The locus identification number (Locus ID), the gene symbol (gene) and spelled out designation of each gene are shown. For the sake of simplicity, EST clones with unknown functions have not been included in the list. simplicity, EST clones with unknown functions have not been included in the list.

 $b_{\mbox{Fold change in gene expression levels upon a 24-h-long treatment with Y27632.}}$ *b*Fold change in gene expression levels upon a 24-h-long treatment with Y27632.

 ϵ p-values of genes affected in NIH3T3 cells determined with the F-statistic. p-values lower than 1.00E-04 are shown as <1.00E-04 *c*p-values of genes affected in NIH3T3 cells determined with the F-statistic. *p*-values lower than 1.00E-04 are shown as <1.00E-04