Retrospective screening of microarray data to identify candidate IFN-inducible genes in a HTLV-1 transformed model

ALAA REFAAT¹ ², MOHAMED OWIS³, SHERIF ABDELHAMED⁴, IKUO SAIKI⁴ and HIROAKI SAKURAI²

¹Drug Resistance Group, Centre for Cancer Research and Cell Biology, School of Medicine, Dentistry and Biomedical Science, Queen's University Belfast, Belfast BT9 7AE, Northern Ireland; ²Department of Cancer Cell Biology, Graduate School of Medical and Pharmaceutical Sciences, University of Toyama, Toyama 930-0194, Japan; ³Systems and Biomedical Engineering Department, Faculty of Engineering, Cairo University, Giza 12613, Egypt; ⁴Division of Pathogenic Biochemistry, Institute of Natural Medicine, Graduate School of Medical and Pharmaceutical Sciences, University of Toyama, Toyama 930-0194, Japan

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Abstract. HuT-102 cells are considered one of the most representable human T-lymphotropic virus 1 (HTLV-1)-infected cell lines for studying adult T-cell lymphoma (ATL). In our previous studies, genome-wide screening was performed using the GeneChip system with Human Genome Array U133 Plus 2.0 for transforming growth factor-β-activated kinase 1 (TAK1)-, interferon regulatory factor 3 (IRF3)- and IRF4-regulated genes to demonstrate the effects of interferon-inducible genes in HuT-102 cells. Our previous findings demonstrated that TAK1 induced interferon inducible genes via an IRF3-dependent pathway and that IRF4 has a counter-acting effect. As our previous data was performed by manual selection of common interferon-related genes mentioned in the literature, there has been some obscure genes that have not been considered. In an attempt to maximize the outcome of those microarrays, the present study reanalyzed the data collected in previous studies through a set of computational rules implemented using ‘R’ software, to identify important candidate genes that have been missed in the previous two studies. The final list obtained consisted of ten genes that are highly recommend as potential candidate for therapies targeting the HTLV-1 infected cancer cells. Those genes are ATM, CFTR, MUC4, PARP14, QK1, UBR2, CLEC7A (Dectin-1), L3MBTL, SEC24D and TMEM140. Notably, PARP14 has gained increased attention as a promising target in cancer cells.

Introduction

In 1977, a group of scientists from Japan identified adult T-cell lymphoma (ATL) as a helper T-cell malignancy (1). A year later, HuT-102 cells were developed from a patient with cutaneous T-cell Lymphoma (2). HuT-102 cells were identified as a subset of helper T-cells with high expression of IL-2 receptor and constitutive IL-2 production, aiding in an autonomous proliferation (2). In 1980, the human T-lymphotropic virus 1 (HTLV-1) was identified in HuT-102 cells, followed by its discovery in cells from ATL patient. Analysis of serum samples from around the world provided information of endemic areas for HTLV-1, and its role as an etiologic agent for ATL (1).

Similar to freshly obtained cells from ATL patients, HuT-102 cells is a helper T-cell cell that express Tax protein, which promote these cells particularly as an excellent model for studying ATL (3). Our previous findings have identified the interferon regulatory factor (IRF) family members as efficient mediators that modulate cytokines profile in those cells, hence providing therapeutic solutions (3). Although the microarray data upon which we based our previous studies were used to elaborately cover the roles of IRFs, particularly IRF3 and IRF4, on Interferon-inducible genes, it was performed by manual selection of related genes mentioned in the literature (3,4). Therefore, we try in the current study to reanalyze the previous data through more efficient computational tools. The aim of browsing our previous data for further analysis is to find important genes that we might have passed away, in order to discover promising opportunities against ATL.

Materials and methods

Gene expression was analyzed using a GeneChip system with Human Genome Array U133 Plus2.0 (Affymetrix; Thermo Fisher Scientific, Inc., Waltham, MA, USA) as described previously (3,4). The arrays were performed independently in two studies as follow, a total of four arrays for the first study: Two for HuT102-shLuc cells and two for HuT102-shTAK1

Correspondence to: Dr Alaa Refaat, Drug Resistance Group, Centre for Cancer Research and Cell Biology, School of Medicine, Dentistry and Biomedical Science, Queen's University Belfast, 97 Lisburn Road, Belfast BT9 7AE, Northern Ireland
E-mail: refaat.alaa@gmail.com

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cells (deposited in the GEO database under accession no. GSE16219). In the second study, six arrays were used: Two for HuT102-siLuc cells, two for HuT102-siIRF4 cells, and two for HuT102-siIRF3 cells (deposited in the GEO Database under accession no. 22036). The findings of both studies are summarized in Fig. 1. In order to better analyze our study, we restricted the number of the analyzed genes. Our criteria was to focus on the genes which have markedly changed. We primarily excluded genes with values between 0.60 and 1.5 fold changes. Genes equal to or less than 0.60 are considered downregulated, and genes equal to or more than 1.50 are considered upregulated. The number of genes remaining after this step was 9,302, 1,343, and 1,447 genes for TAK1, IRF3, and IRF4 arrays, respectively. Next, we used GeneVenn (genevenn.sourceforge.net) to generate a Venn chart demonstrating the set of common genes between the different sets of arrays. The Venn chart indicated a list of 706 common genes between TAK1 and IRF3, 751 common genes between TAK1 and IRF4, 633 common genes between IRF3 and IRF4, and finally 346 common genes between the three datasets. Finally, R software for statistical computing of data sets was used, as explained below, to restrict our selection to the most promising genes, as indicated.

**Results**

*Rules applied on the identified datasets.* In this study, we focused on IRF4 compared to both TAK1 and IRF3, or IRF3
only. Therefore, we continued our analysis using either the 346 common genes between the three datasets (Set A), or the 633 common genes between IRF3 and IRF4 (Set B) (Fig. 2).

To filter down the list of genes generated from the Venn chart, we applied simple rules for the inclusion of genes. We firstly denoted the basic level of a given gene as the average of luciferase transfected sample in the given arrays. We then identified the basic level with different class categories, and applied rules of inclusion for genes according to Tables I and II. Through the applied criteria, we obtained 85 genes out of Set A, and 108 genes out of Set B (Figs. 3 and 4). The next stage was to identify genes inversely correlated between IRF4 against both TAK1 and IRF3, and IRF4 against IRF3 (Fig. 4). From this step, we obtained 14 genes from Set A, and 16 genes from Set B. As our goal from this analysis is to identify new genes inversely controlled by the IRF4 axis vs. the TAK1-IRF3 axis, we excluded unspecified genes, and genes related to interferon for being already covered in previous studies (3, 4) (Tables III and IV, and Fig. 4).

Table I. Rules implemented for set A (IRF4#IRF3/TAK1) and the perspective number of genes.

| Basic level | Class | No. of genes | Rule          | No. of genes after applying the rule |
|-------------|-------|--------------|---------------|-------------------------------------|
| >100        | 5     | 13           | Include all   | 13                                  |
| 50-100      | 4     | 18           | Include if 2 arrays are below 0.5 or more than 2 | 15                                  |
| 25-50       | 3     | 94           | Include if 2 arrays are below 0.4 or more than 4 | 32                                  |
| 10-25       | 2     | 132          | Include if 2 arrays are below 0.3 or more than 6 | 21                                  |
| 0-10        | 1     | 89           | Include if 2 arrays are below 0.2 or more than 8 | 4                                   |
| Total       |       | 346          |               | 85                                  |

| IRF, interferon regulatory factor |

TAK1, transforming growth factor-β-activated kinase 1; IRF, interferon regulatory factor.

Table II. Rules implemented for set B (IRF4#IRF3) and the perspective number of genes.

| Basic level | Class | No. of genes | Rule          | No. of genes after applying the rule |
|-------------|-------|--------------|---------------|-------------------------------------|
| >100        | 5     | 13           | Include all   | 13                                  |
| 50-100      | 4     | 72           | Include if 2 arrays are below 0.5 or more than 2 | 49                                  |
| 25-50       | 3     | 186          | Include if 2 arrays are below 0.4 or more than 4 | 38                                  |
| 10-25       | 2     | 201          | Include if 2 arrays are below 0.3 or more than 6 | 8                                   |
| 0-10        | 1     | 161          | Include if 2 arrays are below 0.2 or more than 8 | 0                                   |
| Total       |       | 633          |               | 108                                 |

| IRF, interferon regulatory factor |

i) Ataxia telangiectasia mutated (ATM) is a Serine/threonine protein kinase with a distinct role in double-strand breaks. Its activation causes the phosphorylation and the activation of several downstream DNA damage and cell cycle arrest checkpoints including the histone variant H2ax, the effector protein kinases, and the tumor suppressor p53 (5-7). Germ-line mutations of ATM lead to ataxia-telangiectasia and shows a high risk of breast cancer (8).

ii) Cystic Fibrosis Transmembrane conductance Regulator (CFTR) is Chloride channel membrane protein (9). It is a member of the ABC transporter superfamily and is activated by phosphorylation by PKA. CFTR conducts anions to flow down their electrochemical gradient and facilitate the passive movement of the positively charged ions. The CFTR gene has been reported to be mutated in patients with cystic fibrosis (CF) and has been a potential therapeutic target. In 2012, the FDA approved Ivacaftor as the first targeted therapy for patient with CF (10). Recently CFTR was identified as a tumor suppressor including intestinal, prostate and ovarian cancer (11-13). Moreover, CFTR was shown as a potent suppressor of epithelial-to-mesenchymal transition (EMT) breast cancer cells and was associated with poor prognosis in patients (14).

iii) Mucin 4 (MUC4) is a high molecular weight glycosylated protein, which plays various roles in promoting cancer progression (15). Overexpression of mucins has been shown to reduce of cell adhesion and promote cancer cell migration and metastasis. MUC4 also phosphorylate the ErbB2 leading to increasing the tumor cell proliferation independent on the activation of neither the MAPK nor AKT pathways. Importantly, MUC4 has been shown as a promising biomarker for diagnoses of pancreatic cancer with undetectable levels in normal pancreas (16).
iv) Poly (ADP-ribose) polymerase 14 (PARP14) is a member of the PARP family of proteins, which involves in DNA repair and programmed cell death. PARP14 was shown to regulate the STAT6-dependent transcription (17). Moreover, Iansante et al has reported the vital role of PARP14 to mechanistically link apoptosis to metabolism. Through inhibiting the JNK1/PKMa2 regulatory axis, PARP14 potentially promoted Warburg effect in hepatocellular carcinoma (18).

v) Quaking homolog, KH domain RNA binding (QK1) is a RNA-binding protein. It is a member of the Signal Transduction and Activation of RNA (STAR) proteins family. QK1 was reported to play a distinct role in schizophrenia via regulating the myelin-related genes activity (19). QKI has been recently shown to be downregulated in lung cancer resulted in poorer prognosis (20). Additionally, it suppresses glioblastoma by stabilizing microRNA-20a leading to regulating TGFβ pathway (21).

vi) UBR2 is an E3 ubiquitin-protein ligase, which acts by recognizing and binding to the N-terminal residues-carrying proteins leading to their ubiquitination and degradation. Notably, UBR2 together with UBR1 were shown as potential negative regulators of mammalian target of rapamycin (mTOR) pathway via degrading the Leucine protein, and therefore its expression can be associated with cancer elimination (22).

vii) C-type lectin domain family 7 member A (CLEC7A, also known as Dectin-1) is a glycoprotein with a distinct role in regulating innate immunity. It is predominantly expressed in macrophages, neutrophils and dendritic cells and binds both CD4+ and CD8+ T cells (23). Dectin-1 also recognizes several...
fungal species, which triggers the induction of numerous cytokines and chemokines including TNF-α, IL-2, IL-6, and IL-23 (24). Mechanistically, Dectin-1 ligand has been shown to act via recruiting and activating the NFκB inflammatory pathway (25). Importantly, activation of dectin-1 on macrophages induces an adaptive immune suppression and promotes pancreatic cancer progression (26).

viii) Lethal (3) malignant brain tumor-like protein (L3MBTL) is a polycomb group protein (PcG) that recognizes and binds methylation residues on the target proteins leading to post-translational modifications. L3MBTL was shown to be associated with myeloid malignancies (27).

ix) SEC24D is a member of the SEC23/SEC24 family, which is involved in vesicle trafficking at the endoplasmic reticulum (28).

x) TMEM140 is the transmembrane protein 140, shown as a promising prognostic marker for patients with glioma where its overexpression strongly correlates with tumor size.

| Gene symbol | Gene name | Array 1 (TAK1/luc) | Array 2 (IRF3/luc) | Array 3 (IRF4/luc) | Class |
|-------------|-----------|--------------------|--------------------|--------------------|-------|
| ATM         | Ataxia telangiectasia mutated | 0.52               | 0.26               | 12.87              | 2     |
| CFTR        | Cystic fibrosis transmembrane conductance regulator | 4.31               | 2.11               | 0.34               | 3     |
| CXCL10      | Chemokine (C-X-C Motif) ligand 10 | 0.35               | 0.19               | 6.55               | 5     |
| CXCL11      | Chemokine (C-X-C Motif) ligand 11 | 0.21               | 0.19               | 8.98               | 5     |
| FLJ101213   | Hypothetical protein LOC55096 | 0.58               | 0.16               | 4.23               | 3     |
| GBP4        | Guanylate binding protein 4 | 0.51               | 0.45               | 7.51               | 4     |
| IFIT1       | Interferon-induced protein with tetratricopeptide repeats 1 | 0.36               | 0.33               | 5.27               | 5     |
| IFIT2       | Interferon-induced protein with tetratricopeptide repeats 2 | 0.5               | 0.29               | 2.49               | 5     |
| IFIT3       | Interferon-induced protein with tetratricopeptide repeats 3 | 0.48               | 0.43               | 2.23               | 5     |
| IL18RAP     | Interleukin 18 receptor accessory protein | 0.22               | 0.49               | 2.21               | 5     |
| MUC4        | Mucin 4, cell surface associated | 1.66               | 3.29               | 0.31               | 3     |
| PARP14      | Poly (ADP-ribose) polymerase family, member 14 | 0.1               | 0.58               | 2.94               | 5     |
| QKI         | QKI, KH domain containing, RNA binding | 4.38               | 5.4               | 0.59               | 2     |
| UBR2        | Ubiquitin protein ligase E3 component N-recognin 2 | 0.29               | 0.23               | 1.94               | 2     |

**Table III. Final list of inversely correlated genes created from set A (IRF4#IRF3/TAK1).**

**TAK1, transforming growth factor-β-activated kinase 1; IRF, interferon regulatory factor.**

| Gene symbol | Gene name | Fold 1 (IRF3/luc) | Fold 2 (IRF4/luc) | Class |
|-------------|-----------|-------------------|-------------------|-------|
| ATM         | Ataxia telangiectasia mutated | 0.26               | 12.87              | 3     |
| CLEC7A      | C-type lectin domain family 7, member A | 0.14               | 4.62               | 2     |
| CXCL10      | Chemokine (C-X-C Motif) ligand 10 | 0.19               | 6.55               | 5     |
| CXCL11      | Chemokine (C-X-C Motif) ligand 11 | 0.19               | 8.98               | 5     |
| FLJ101213   | (unidentified) | 0.16               | 4.23               | 3     |
| GBP4        | Guanylate binding protein 4 | 0.45               | 7.51               | 4     |
| IFIT1       | Interferon-induced protein with tetratricopeptide repeats 1 | 0.33               | 5.27               | 5     |
| IFIT2       | Interferon-induced protein with tetratricopeptide repeats 2 | 0.29               | 2.49               | 5     |
| IFIT3       | Interferon-induced protein with tetratricopeptide repeats 3 | 0.43               | 2.23               | 5     |
| IL18RAP     | Interleukin 18 receptor accessory protein | 0.49               | 2.21               | 5     |
| ISG20       | Interferon stimulated exonuclease gene 20 kDa | 0.45               | 2.24               | 5     |
| L3MBTL      | Lethal(3)malignant brain tumor-like gene | 0.13               | 4.84               | 3     |
| LOC729397   | Hypothetical LOC729397 | 0.31               | 2.23               | 4     |
| PML         | Promyelocytic leukemia | 0.53               | 2.06               | 5     |
| SEC24D      | SEC24 family member D | 0.31               | 3.24               | 3     |
| TMEM140     | Transmembrane protein 140 | 0.35               | 2.61               | 4     |

**Table IV. Final list of inversely correlated genes created from set B (IRF4#IRF3).**

**IRF, interferon regulatory factor.**
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