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Microarray analysis reveals potential mechanisms of *BRMS1*-mediated metastasis suppression

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Abstract

We used Affymetrix oligonucleotide microarrays to compare gene expression profiles of the metastatic parental breast cancer cell line MDA-MB-435 (435) and the non-metastatic daughter cell line created by the stable expression of the Breast cancer Metastasis Suppressor 1 (*BRMS1*) gene in 435 cells, MDA-MB-435-BRMS1 (435/BRMS1). Analysis of microarray data provided insight into some of the potential mechanisms by which *BRMS1* inhibits tumor formation at secondary sites. Furthermore, due to the importance of the microenvironment, we also examined gene expression under different growth conditions (i.e., plus or minus serum), however, expression of 565 genes was significantly (adjusted *p*-value <0.05) altered regardless of *in vitro* growth conditions. *BRMS1* expression significantly increased multiple major histocompatibility complex (MHC) genes and significantly decreased expression of several genes associated with protein localization and secretion. The pattern of gene expression associated with *BRMS1* expression suggests that metastasis suppression may be mediated by enhanced immune recognition, altered transport, and/or secretion of metastasis-associated proteins.

Keywords

Affymetrix; breast cancer; *BRMS1*; metastasis; microarray; MHC

Introduction

Breast cancer metastasis suppressor 1 (*BRMS1*) is one of a growing family of genes termed metastasis suppressors, that block development of metastasis without preventing orthotopic tumor growth [1-3]. For the most part, the mechanisms by which metastasis suppressors inhibit are not well understood. *BRMS1* induces several apparently unrelated phenotypic changes – restoration of homotypic gap junctional intercellular communications [4,5], inhibition of NFκB signaling [6], abrogation of phosphoinositide signaling [7].

The breadth of *BRMS1*-induced cellular changes are, in part, explained by its role as a component of core mSin3a:histone deacetylase complexes and interaction with the retinoblastoma tumor suppressor [8-10]. Thus, *BRMS1* is positioned within the cell to regulate

expression of a plethora of genes, yet histone deacetylase inhibitors (and by inference, the acetylases and deacetylases themselves) regulate only approximately 2% of known genes [11]. The objective of the current study was to compare gene expression patterns in BRMS1-expressing vs. non-expressing human breast carcinoma cells. We hypothesized that insights regarding BRMS1-associated changes, in particular, and perhaps cancer metastasis in general, would be gained by using unbiased genome-wide expression analyses. Our data shows that BRMS1 expression significantly altered genes involved in: (1) immune response; (2) protein transport and localization; and, (3) Golgi structure and function.

Materials and Methods

Cell Culture and Harvest for RNA isolation

Cells were cultured in Corning™ T-175 flasks (Corning catalog no. 431080, Fisher Scientific, Santa Clara, CA) in HyQ DMEM high glucose/SF-12 medium (HyClone, Logan, UT) supplemented with 5% fetal bovine serum (FBS) (Atlanta Biologicals, Lawrenceville, GA), 1% nonessential amino acids, 1.0 mM sodium pyruvate and maintained at 37°C with 5% CO₂ in a humidified atmosphere. Cells were passaged using 0.125% trypsin for the parental 435 cells or 2.0 mM EDTA for the daughter 435/BRMS1 cells in Ca²⁺/Mg²⁺-free Dulbecco's phosphate buffered saline.

After attachment cells were washed with 1× phosphate buffered saline (PBS). Medium was then replaced for one hour with either fresh complete media containing 10% FBS or complete serum-free medium. After one hour medium was removed and cells were harvested by washing once with 1× PBS, followed by addition of 10 mL of Trizol® (Invitrogen, Carlsbad, CA). Trizol® was incubated with cells for 3-5 minutes at room temperature before thorough rinsing of flask bottom and removal into a 15 mL conical tube. Samples were then frozen at -80°C at least overnight prior to isolation of total RNA. Samples were thawed at room temperature for 60 minutes. RNA isolation was done by following the Invitrogen protocol except centrifugation steps were done at 3,200 × g for 30 minutes or 10 minutes at 2 to 8°C in place the “long” or “short” centrifugation steps in the protocol.

Microarray Hybridizations and Analyses

Total RNA was submitted to the Utah State University Affymetrix Core Facility for labeling and hybridization. Samples were assessed for quantity on a NanoDrop ND 1000 spectrophotometer and quality using an Agilent (Santa Clara, CA) Bioanalyzer 2100. Samples were labeled using the Affymetrix GeneChip® One-Cycle Target Labeling and Control Kit (part # 900493) as described in the Affymetrix Eukaryotic RNA One-cycle cDNA Synthesis and Labeling manual using an MJ Research DNA Engine Tetrad 2 thermocycler® (Bio-Rad Laboratories, Hercules, CA). Hybridization was performed using a GeneChip Hybridization Oven 640® (Affymetrix, Santa Clara, CA). Washing and staining was performed using an Affymetrix GeneChip Fluidics Station 650® and chips were scanned using an Affymetrix GeneChip Scanner 3000® and images interpreted and monitored using GCOS version 1.2.

Briefly, single- then double-stranded cDNA was synthesized using a T7-oligo primer followed by sample purification. Synthesis of biotin-labeled cRNA proceeded using clean double-stranded cDNA and the Affymetrix IVT labeling kit. Labeled cRNA was purified, quantified and fragmented prior to addition to pre-warmed U133 +2.0 chips. Following overnight hybridization, chips were washed and subjected to stain solution, then washed to remove excess stain. After scanning, images were assessed for quality using default metric settings. Both raw and quality controlled data were received for each sample.

Data was analyzed by standard methods using R [12] and Bioconductor [13] (open source program; <http://www.bioconductor.org>) for all statistical analyses. The eight Affymetrix Gene Chip array raw data files were normalized using the robust multi-array analysis (RMA) procedure in Bioconductor's affy package [14], which includes background correction, quantile normalization, and probe set summary. Using Bioconductor's limma package [15], a linear model was then fit to the normalized data and contrasts tested for a cell line-media interaction and cell line and media main effects. The top 750 significant cell line effect probe sets were selected to generate a working gene list, which corresponded to an FDR-adjusted p-value cutoff of $p < 0.05$ and B (log odds) cutoff of 0.

The working list of significant probe sets was used as the basis for performing a functional over-representation analysis using mappings to the Gene Ontology (GO) [16]. The working list of significant probe sets was first converted to a list of unique Entrez Gene [17] identifiers. Probe sets were then matched to Entrez Gene entries, and the genes were mapped to their representative GO identifiers, as well as all connected GO ancestors. Using a hypergeometric p-value, each GO identifier was evaluated for over-representation in the list of significant genes. Due to the nature of the hypergeometric calculation and its testing of many dependent hypotheses, interpretation of this p-value is not straightforward. In addition, GO identifiers with few annotated genes tend to show artificially small p-values. We therefore used the developer's suggested criteria and focused our attention on GO identifiers having a small p-value ($p < 0.05$) and at least 10 member genes as significantly over-represented [18].

The expression patterns of genes found in selected over-represented GO terms were plotted on heat maps with the gplots package [19] using Euclidean hierarchical clustering on the genes. Replicates and multiple probe sets when present for a given gene were averaged for display purposes.

qRT-PCR

Total RNA was thawed on ice and aliquots were diluted to provide standards at 100, 50, 10, 0.1 and 0.05 nanograms (ng) final concentration. The hypoxanthine guanine phosphoribosyl transferase (HPRT) gene was used as the standard. Samples were probed for relative copy number using a start concentration of 10 ng of total RNA. Samples were prepared using TaqMan primers and probe sets (GeneScript, Inc., Scotch Plains, NJ) and the Qiagen (Valencia, CA) QuantiTect Probe RT-PCR kit according to the directions. Samples were processed and recorded using a M.J. Research Opticon 2 system. Data analyses were performed using Opticon 2 software.

RESULTS

The experimental and analytical approaches used in this study have been outlined in a flow chart (appendix, Fig 1). RNA was isolated from 435 and 435/BRMS1 cells cultured for one hour under serum-containing or serum-free growth conditions, quantified and hybridized to Affymetrix human U133 plus 2.0 chips. Normalization and analysis of the raw data files from the microarray analyses was performed (see Materials and Methods) and a linear model was then fit to the normalized data and contrasts tested for cell line or media effects. No media effects were identified (effects specific to the presence or absence of serum), but some gene chip probe sets were found to have a significant (adj. p -value <0.05) cell line effect. The top 750 significant cell line effect probe sets were selected, and the redundant probe sets for specific genes were removed to produce a working gene list, which corresponded to an FDR-adjusted p-value cutoff of $p < 0.05$ and B (log odds) cutoff of 0.

The differentially expressed gene list generated using this approach reveals patterns or groups of genes whose expression is changed as a result of BRMS1 expression in the 435 cell line.

From our differentially expressed gene list the 40 genes whose expression showed the greatest fold change induction were compiled (Table 1). This list contains genes in descending order by fold change that have been implicated in the immune response, signal transduction, metabolism, transcription, cytoskeleton, Golgi apparatus, and transport. The 40 genes whose expression showed the greatest fold change repression are shown in Table 2. These genes have been associated with transport, cell adhesion, signal transduction, metabolism, cell proliferation, cytoskeleton, and extracellular matrix.

Our gene lists provide some insight into the effect of BRMS1 expression, but information in this format fails to uncover the overall biological relevance of the transcriptional changes that lead to a block in metastasis resulting from BRMS1 expression. A statistically sound method for summarizing significant gene lists into the biological processes they represent has been developed by Gentleman [18]. This technique, known as over-representation analysis, was applied to the data.

The working list of significant probe sets was used as the basis for performing a functional over-representation analysis using mappings to Gene Ontology (GO). This technique helps to highlight the overall processes that our list of significant genes represents, and offers an effective method to reveal the biological relevance of transcriptional changes which may not be apparent examining the list on a gene-by-gene basis.

The results of functional over-representation analysis are shown as directed acyclic graphs (DAG) with each graph node representing a GO term. GO terms found to be significantly over-represented were colored according to the collective trend for induction or repression among the genes at that GO term. Red indicates that genes at that node show collective mean induction in 435/BRMS1 cells, while blue indicates collective mean repression. Nodes not colored are not significantly over-represented, but are included to maintain connectivity. Significantly over-represented GO terms within the biological process (BP) GO category are shown (Fig. 1, A). Some of the most highly over-represented BP GO terms include immune response (Fig. 1, B) and protein localization (Fig. 1, C). For the cellular component (CC) GO category (data not shown), the most notable over-representation was found at the Golgi apparatus.

To elucidate which nodes contribute to over-representation within the general GO term of immune response an expansion of the DAG of BP GO terms with both ancestor and child terms is shown in more detail (Fig. 2, A). The function depicted in the expansion proceeds from general in the bottom nodes to very specific in the top nodes (Fig 2, A). To visualize expression of individual genes expanded heat maps were created for specific BP GO terms (appendix Fig. 2, 3, 4). Figure 2 B depicts a combination figure of DAG representation for individual genes which fall within child nodes of the “immune response” node (j in Fig. 2, A) combined with the heat map expression intensities for each.

The full heat map of the expression of the genes within the BP GO term immune response is shown (appendix Fig. 2). A more detailed list of genes significantly (adj. p -value <0.05) induced within the immune response BP GO term is shown in tabular form in the appendix (Table 1, appendix). Of the genes most highly induced in 435/BRMS1 cells are several encoding major histocompatibility complex (MHC), class I and II proteins, including: HLA-DQB1, HLA-DRB1, HLA-DRB5, HLA-DMB, HLA-DQA1, HLA-DPA1, HLA-DRA, HLA-DRB4, HLA-DMA, C1S, HLA-B, HLA-C, and HLA-F.

Our analyses also uncovered changes in the expression of genes associated with protein localization/transport (Fig. 2, A and B). Significant (adj. p -value <0.05) genes were found to be associated with the BP GO term protein localization, as well as child GO terms intracellular protein transport, protein secretion, protein transport, and establishment of protein localization (Fig. 2, A). A more detailed expression profile for individual genes identified in the protein

localization BP GO term is shown as a node delineated heat map in Figure 4 (B) and in the full heat map (appendix Fig. 3).

For the protein localization GO term, expression of several genes is repressed coincident with BRMS1 expression including: RBP1, RAB1A, APBA2BP, TIMM22, DNAJC1, RAB8B, SNX1, SSR3, SNX5, and RAB27A. Genes whose expression is induced as a result of BRMS1 expression include: SEC23IP, SEC3L1, SYTL2, COG1, NUP98, COPZ2, CD74 (2), AP1S2, A2M, SCG2, FYB, AP1G2, TGFB1, TLK1, GNAS, and PEX13.

The genes significantly (adj. p -value <0.05) altered in 435/BRMS1 cells identified by over-representational analysis to be involved in the Golgi apparatus GO node under Cellular Component are shown as a heat map (appendix Fig. 4). Those whose expression is significantly repressed coincident with BRMS1 expression include: D4ST1, PIK4CA, ATP7A, ST6GALNAC2, APBA2BP, RAB1A, B3GNT1, LARGE, GOLGA8A, AP1G2, ST6GALT1, SNX1, and B4GALT1. Genes whose expression is significantly induced as a result of BRMS1 expression include: COG1, SEC23IP, PSENN, AP1S2, COPZ2, BACE1, BICD1, VAMP4, IL15, and HS3ST3A1.

In an effort to test the validity of the microarray data that we obtained, we used quantitative RT-PCR as a second technique to examine changes in gene expression. Five genes from our differentially expressed gene list were selected at random for confirmation using real time quantitative PCR (qRT-PCR). TaqMan primers and probes were obtained (GeneScript or Applied Biosystems, Inc.) for *BRMS1*, Endothelin converting enzyme 2 (*Ece2*), Ectonucleotide pyrophosphatase phosphodiesterase 2 (*Enpp2*), Abl-interacting protein 2 (*Abi2*) and Caspase 1 (*Casp1*). qRT-PCR was performed using a one-step protocol. Under our experimental conditions only amplification of *Ece2* produced signal below our lowest control making a reasonable assessment of its expression level in either cell type inconclusive. All other products from amplification (Fig 5, appendix) clearly corroborated our analyses of the microarray data.

Discussion

Metastasis is a complex, multi-step event involving cellular processes such as protein transport, cytoskeletal rearrangements and signal transduction. The metastatic process is sequential and inhibition of any step precludes completion of subsequent steps. Re-expression of BRMS1 prevents metastasis but the step(s) at which BRMS1 exerts its effects are still not completely understood. Preliminary data suggest that BRMS1 controls as multiple steps, including outgrowth at the secondary site (P.A. Phadka and D. Welch, unpublished).

While the amino acid sequence of BRMS1 provided minimal clues regarding function, identification of proteins with which BRMS1 interacts led to the understanding that BRMS1 is involved in transcriptional machinery. Specifically, the BRMS1 protein is a component of multiple mSIN3a:histone deacetylase complexes. Because these complexes regulate chromatin structure, they have the potential to influence gene expression throughout the genome. To better understand BRMS1 function, in particular, its role in regulating gene expression, a genome-wide unbiased expression study was initiated.

Using Affymetrix platforms coupled with robust informatic analyses, we found that BRMS1 re-expression selectively changed expression of certain categories of genes, especially MHC Class I and II and molecules involved in protein transport and secretion. The patterns of gene expression change provide insights into the steps of metastasis influenced by BRMS1 expression.

The most dramatically over-represented gene ontology designation was host immune system, which is a complicated compilation of molecules and events involved in immune response

(Table 1, Fig. 2; appendix Table 1, Fig. 2). In response to BRMS1 expression, the expression of several immune response genes was altered including: MHC class I, MHC class II, interleukins, interferon-induced, interferon-induced protein with tetratricopeptide repeats, CD antigens and several with roles in processing or transport of antigens.

Previous studies have shown that some MHC class I genes are not expressed in certain cancers, which may alter the immune response [20-22]. Altered expression of MHC, class I and II molecules has been described in metastatic lesions from breast tumors [23-25] and breast cancer lines [26]. Notably, it has been established that the expression of MHC class I molecules on cancer cells has an important role in the cytotoxic T-lymphocyte (CTL)-mediated immune response [27], and up-regulation of MHC class I genes could provide opportunities for CTL's to recognize and destroy cancer cells [28]. However, a cause and effect relationship has not been clearly established. On a related note, the tumor suppressor pRb appears to be required for up-regulation of some MHC, class II genes [29,30]. Therefore, some similarity may exist in the mechanism of action of tumor suppressors and metastasis suppressors.

MHC, class II classical molecules, HLA-DPA, -DPB, -DQA, -DQB, -DRA and -DRB, are normally expressed only on professional Antigen Presenting Cells (APC'S) and a few other types of cells. These molecules present peptides derived from proteins degraded in endosomes. Earlier studies suggest that changes in expression of MHC, class II molecules may influence tumor cell evasion of immune surveillance favoring metastatic disease [31,32]. MHC class II genes were found repressed in highly metastatic cells [30]. Induction of MHC class I and II genes may be one mechanism by which 435/BRMS1 cells are kept at low populations, by triggering an immune response that eliminates them in lung and bone.

In addition to immune response gene expression changes, GO analyses revealed two other over-represented GO categories, the BP GO term protein localization and CC GO term Golgi apparatus. Reciprocally, one is dependent on the other—the structure of the Golgi apparatus is dependent upon protein localization, and appropriate protein localization is dependent upon the Golgi apparatus for transportation.

The secretory pathway of eukaryotic cells is a highly complex bidirectional system which includes the endoplasmic reticulum, the cis-, medial-, and trans-Golgi, the late endosome, the lysosome/vacuole, and the plasma membrane. Proteins transported in this elaborate system are first synthesized at the endoplasmic reticulum (ER) membrane where they remain in the membrane or are injected into the lumen of the ER. Proteins packaged into lipid vesicles are then transported in an anterograde direction to more distal sites in the pathway via a series of vesicle budding, docking and fusion events. Alternatively, proteins can be retained in secretory compartments or retrieved to a more proximal compartment via a retrograde trafficking process.

The processes of lipid vesicle fission (breaking off the donor membrane), transport, and fusion (with the acceptor membrane) are complex and involve a number of membrane associated and transmembrane proteins. Uncovered in this study were genes representing membrane associated proteins such as vesicle-associated membrane protein 4 (Vamp4), coatamer protein complex-subunit zeta 2 (COPZ2), sec3L1, sec23IP, adaptor-related protein complex subunits (AP1G2 and AP1S2), as well as those genes encoding proteins known to cycle from the cytosol to a membrane (vesicle, Golgi, endosome, or plasma membrane).

Other gene products known to have a role in regulation of vesicle-mediated protein trafficking that are found localized to membranes or in the cytosol include the small monomeric GTPases Rab1a, Rab8b, and Rab27a which have been implicated in vesicle formation, uncoating and docking [33-35]. Notably, Rab27a and the synaptotagmin-like 2 (SYTL2) protein are involved in melanosome transport [36]. The sorting nexins (SNX1 and SNX5) which are known to bind

phosphoinositides (via PX domains) associate with the retromer complex and regulate protein transport at the endosome [37,38]. Also a regulator of vesicle-mediated protein trafficking and phosphoinositide metabolism is the PtdIns 4-kinase (PIK4CA) with a role in protein transport from the Golgi to the plasma membrane [39,40]. Altogether, the data strongly suggest that the expression of BRMS1 alters exocytic protein trafficking, with changes in genes that encode proteins associated with the ER, Golgi, and endosomes. Importantly, changes in proteins associated with Golgi function would have profound effects on protein transport.

Several genes whose protein products have important roles in protein and lipid modification within the Golgi compartments were also altered. Those genes include glycosyltransferases (e.g. LARGE), galactosyltransferase (B4GALT1), N-acetylglucosaminyltransferase (B3gnt1), sialyltransferase (ST6GALNAC2), and sulfotransferase (D4ST1). The decrease in expression of these genes may reflect an overall reduction in post-translational protein modification prior to transport, or be specific to a subset of proteins and lipids.

Two other cellular attributes that seem to be altered in the BRMS1-expressing cells are the cytoskeletal-related genes and cell-adhesion related genes. Several genes whose expression is changed very likely affect how cell-to-cell and cell-to-extracellular matrix interactions occur. Notably, we also observed significant (adj. p -value < 0.05) gene expression changes in connexin 32 (full data set on web site) that was previously implicated in metastasis [5] and associated with BRMS1 expression.

Altogether, what do these results tell us about the mechanism by which metastasis can be suppressed? The inherently simple design of this study was to express a single metastasis suppressor gene, BRMS1, in order to gain insight into potential transcriptional changes that enable switching a highly metastatic cell line to a non-metastatic cell line. The results strongly suggest that BRMS1 suppression of metastasis in part is accomplished by altering the host immune response, intracellular protein transport and Golgi structure and function. In addition to these potentially coordinated effects, changes in expression of numerous genes involved in signal transduction, cell adhesion, cytoskeletal architecture, cell proliferation and metabolism (see summary gene list on web site www.biosystems.usu.edu/ were observed in BRMS1-expressing cells. The breadth of changes is consistent with the observed biology.

Many of the changes observed in gene expression could potentiate altering of the microenvironment of micrometastases via changes in protein transport, secretion, or extracellular matrix affects. Alternatively, observed changes may alter how the non-metastatic 435/BRMS1 cells are able to respond to the stimuli encountered in the microenvironment. That is, the non-metastatic cells cannot respond to growth factor mediated signaling, alter their cytoskeleton or that their cell adhesion properties are altered compared to the metastatic cells. While this study provides processes and genes that should be evaluated, follow-up studies carefully examining cause and effect of individual genes or sets of genes and how these modulate metastasis are necessary.

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Appendix

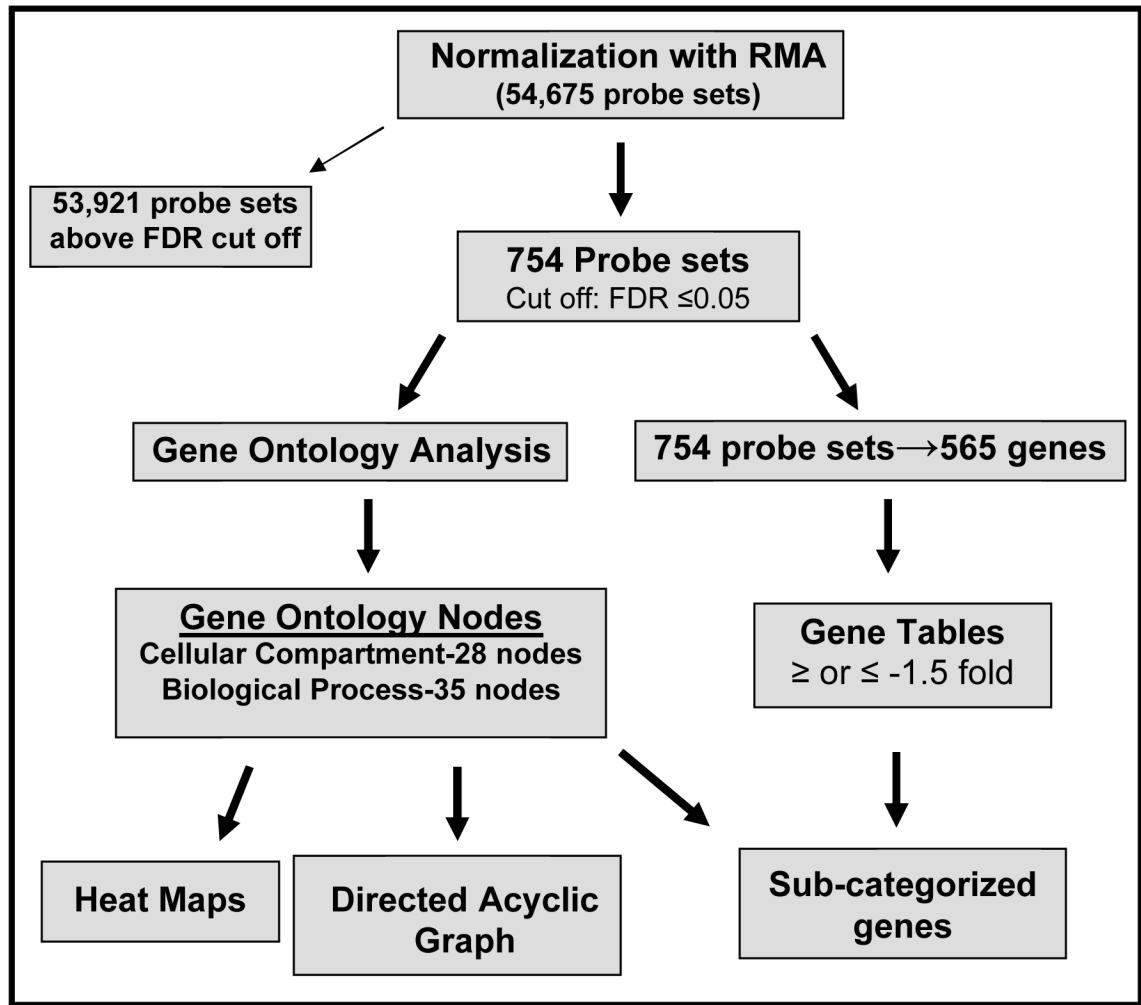


Figure 1.
A flow chart of experimental and analytical approaches.

Appendix

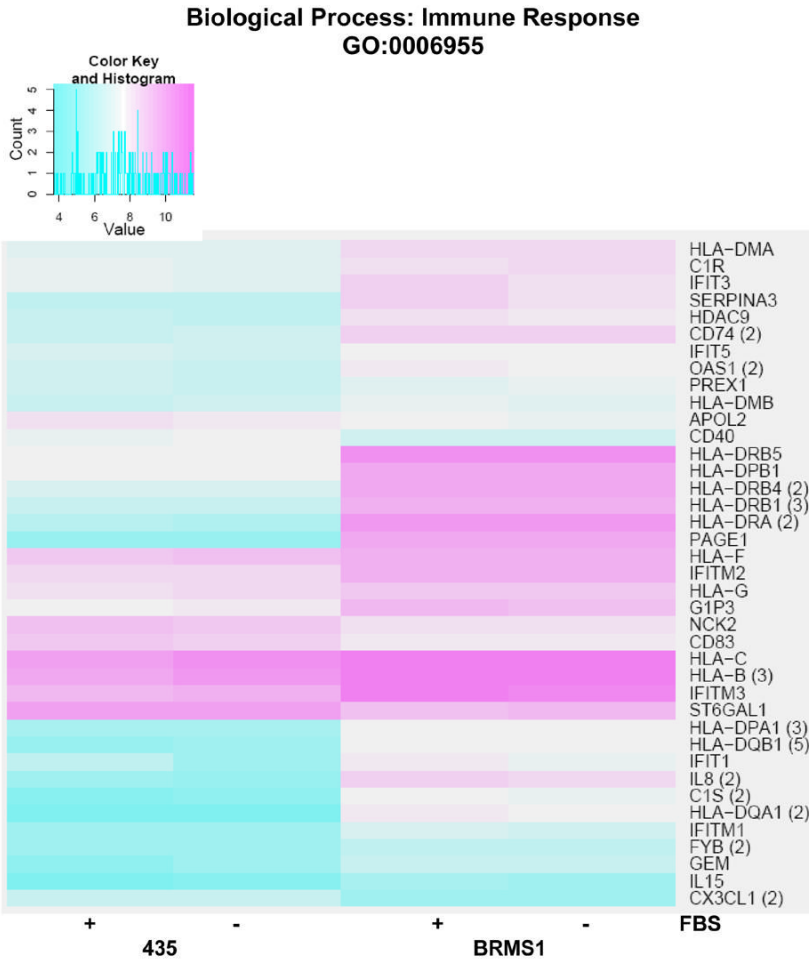


Figure 2. Heat map of significant genes within the biological process (BP) Gene Ontology (GO) term immune response. Color key is shown at top left and represents the \log_2 expression intensity for the average of the probe sets for the specified gene determined to be significantly (adj. p -value < 0.05) altered in our data. Cyan color indicates lower relative expression and magenta indicates higher relative expression. If more than one probe set for a given gene was determined to be significant, they were averaged together prior to heat mapping and the number of probe sets is indicated in parentheses to the right of the gene symbol on the map. The presence or absence of fetal bovine serum (FBS) in the samples from each cell line is indicated at the bottom of the map.

Appendix

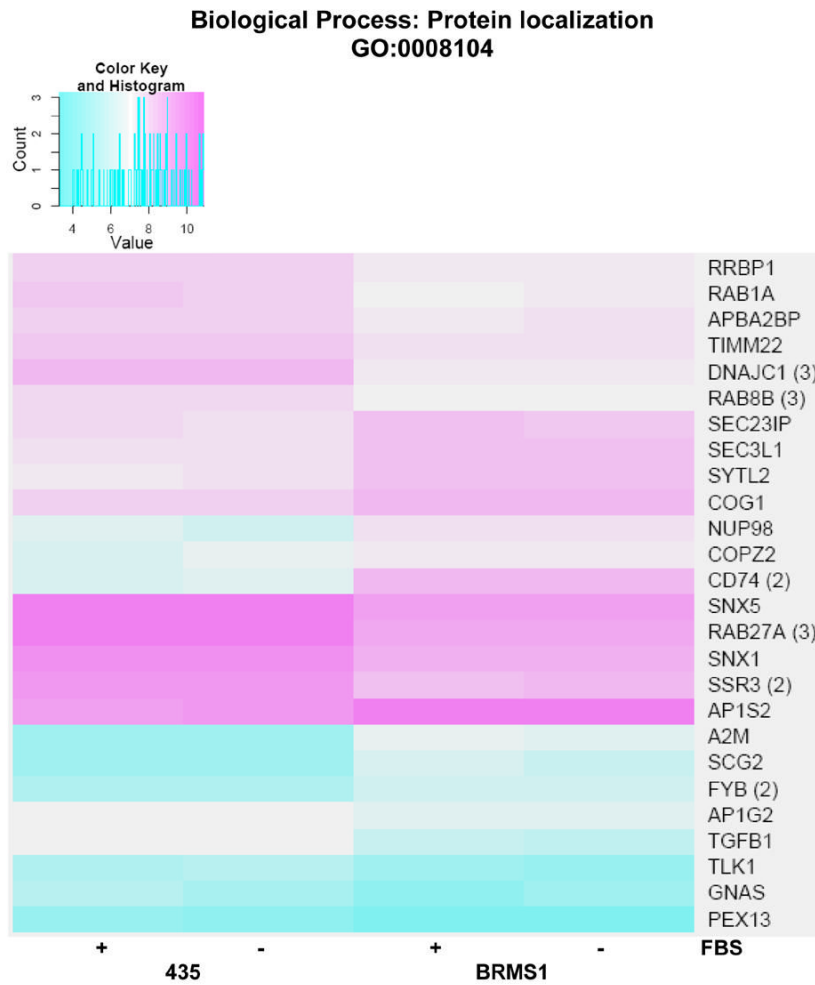


Figure 3. Heat map of significant genes within the biological process (BP) Gene Ontology (GO) term protein localization. Color key is shown at top left and represents the \log_2 expression intensity for the average of the probe sets for the specified gene determined to be significantly altered in our data. Cyan color indicates lower relative expression and magenta indicates higher relative expression. If more than one probe set for a given gene was determined to be significant, they were averaged together prior to heat mapping and the number of probe sets is indicated in parentheses to the right of the gene symbol on the map. The presence or absence of fetal bovine serum (FBS) in the samples from each cell line is indicated at the bottom of the map.

Appendix

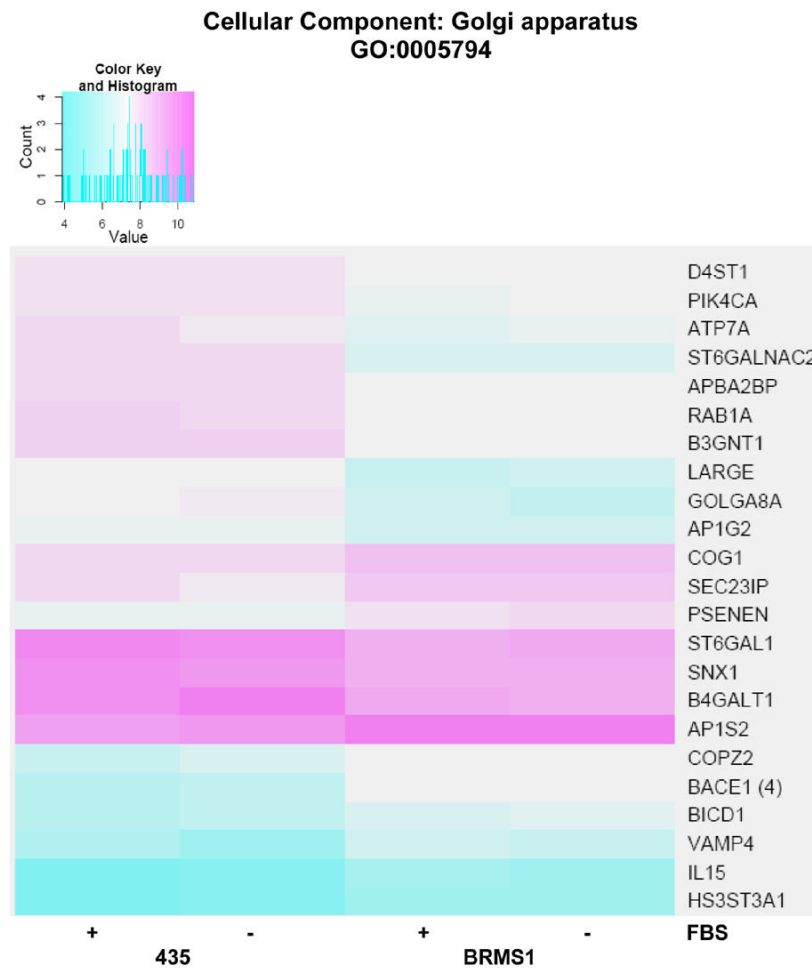
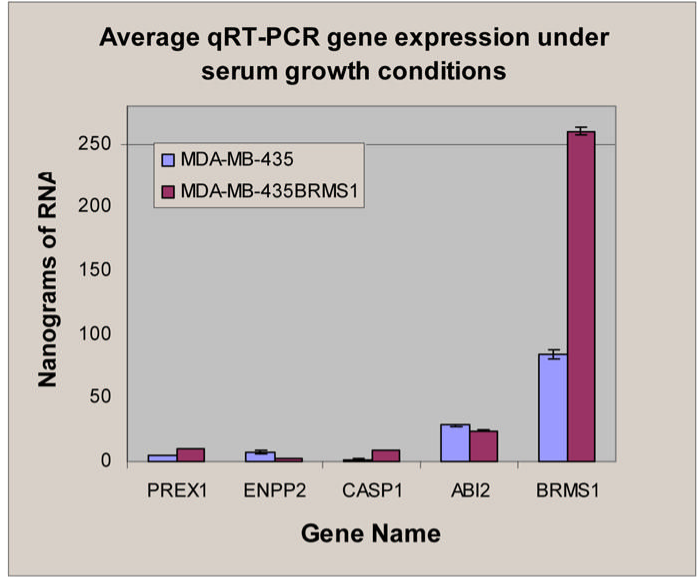


Figure 4. Heat map of significant genes within the cellular component (CC) Gene Ontology (GO) term Golgi apparatus. Color key is shown at top left and represents the log(2) expression intensity for the average of the probe sets for the specified gene determined to be significantly altered in our data. Cyan color indicates lower relative expression and magenta indicates higher relative expression. If more than one probe set for a given gene was determined to be significant, they were averaged together prior to heat mapping and the number of probe sets is indicated in parentheses to the right of the gene symbol on the map. The presence or absence of fetal bovine serum (FBS) in the samples from each cell line is indicated at the bottom of the map.

Appendix

A)



B)

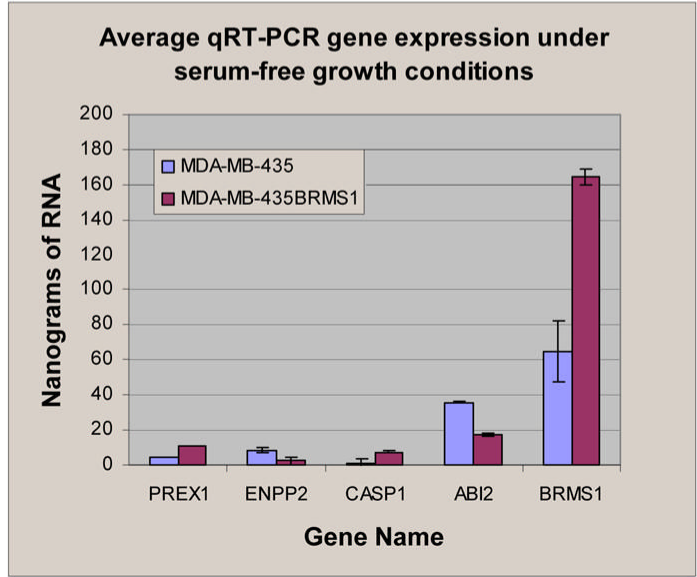


Figure 5. Panel A: Quantitative real-time polymerase chain reaction analysis of select gene expression in nanograms of RNA from 435 and 435/BRMS1 cells grown in serum containing media. Standard deviation of each of the three replicates is shown. Panel B: Quantitative real-time polymerase chain reaction analysis of select gene expression with standard deviation in nanograms of RNA from 435 and 435/BRMS1 cells grown in serum-free media. Standard deviation of each of the three replicates is shown. Gene expression was determined relative to standards from both cell lines using the hypoxanthine guanine phosphoribosyl transferase gene.

Appendix Table 1

Genes involved in immune response induced in 435/BRMS cells relative to 435 cells.

Accession no. ^a	Symbol	Signal Log Ratio ^b	FDR adj P-Value ^c
NM_003785	PAGE1	5.383853	0.000035
NM_019111	HLA-DRA	5.268353	0.000562
NM_002122, 020056	HLA-DQA1/A2	4.639360	0.000035
NM_002124	HLA-DRB1	4.247076	0.000182
NM_000584	IL8	4.117897	0.002229
NM_002123	HLA-DQB1	4.110985	0.000035
NM_001734, 201442	C1S	3.767461	0.003596
NM_002159	HTN1	3.432327	0.017445
NM_002124	HLA-DRB5	3.393892	0.000071
NM_021983	HLA-DRB4	3.367954	0.000075
NM_033554	HLA-DPA1	3.144453	0.000508
NM_004355	CD74	3.033860	0.002536
NM_001187	BAGE	2.777546	0.016837
NM_001085	SERPINA3	2.550661	0.004990
NM_002121	HLA-DPB1	2.393904	0.001461
NM_001472, 001474-77, 021123	GAGE 2, 4-7B	2.259294	0.013588
NM_001001887, 001548	IFIT1	2.146046	0.028883
NM_014707, 058176-77, 178423, 178425	HDAC9	2.109489	0.006050
-----	HLA-DRB6	1.974294	0.002037
NM_005261, 181702	GEM	1.758017	0.005575
NM_002038, 022872, 022873	G1P3	1.682761	0.032661
NM_021034	IFITM3	1.576183	0.017422
NM_003641	IFITM1	1.554440	0.016836
NM_001712	CEACAM1	1.443276	0.016670
NM_006120	HLA-DMA	1.441918	0.015915
NM_006435	IFITM2	1.434174	0.013460
NM_002534, 016816	OAS1	1.432204	0.016408
NM_001011544, 005366	MAGEA11	1.348985	0.046765
NM_005514	HLA-B	1.324459	0.014377
NM_001733	C1R	1.299634	0.022136
NM_001549	IFIT3	1.283534	0.035789
NM_001465, 199335	FYB	1.235783	0.023193
NM_012420	IFIT5	1.125754	0.038872
-----	GAGE3	1.082578	0.029565
NM_000585, 172174, 172175	IL15	0.966510	0.016706
NM_002118	HLA-DMB	0.929749	0.044733
NM_020820	PREX1	0.898315	0.030792
NM_006417	IFI44	0.894730	0.033646
XM_496804	HCG22	0.876532	0.042341
NM_002117	HLA-C	0.850252	0.043950
NM_018950	HLA-F	0.759131	0.041014

^a GenBank accession number

^b Signal Log ratio: power value is 2, for example 5.383853 is equal to $2^{5.383853}$.

^c For explanation of processing used to obtain FDR-adjusted *P*-value see materials and methods.

Appendix Table 2

Genes involved in protein localization/secretion altered in 435/BRMS cells relative to 435 cells.

Accession no. ^a	Symbol	Signal Log Ratio ^b	FDR adj P-Value ^c
Induced			
NM_004355	CD74	3.033860	0.002536
NM_000014	A2M	2.081898	0.004244
NM_003469	SCG2	1.564269	0.002536
NM_005387, 016320, 139131-32	NUP98	1.463201	0.045012
NM_001465, 199335	FYB	1.235783	0.023193
NM_012219	MRAS	1.165204	0.012502
NM_032379, 032943, 206927-30	SYTL2	1.057945	0.005790
NM_016429	COPZ2	1.041162	0.027706
NM_003916	AP1S2	0.997023	0.022876

Accession no. ^a	Symbol	Signal Log Ratio ^b	FDR adj P-Value ^c
NM_018261, 178237	SEC3L1	0.924925	0.040666
NM_006946	SPTBN2	0.805178	0.031828
NM_018714	COG1	0.774169	0.041457
NM_007190	SEC23IP	0.662779	0.041697
Repressed			
NM_003917, 080545	AP1G2	-0.593089	0.044733
NM_031231, 031232	APBA2BP	-0.624727	0.032046
NM_013337	TIMM22	-0.680658	0.033443
NM_012290	TLK1	-0.697219	0.029565
NM_006912	RIT1	-0.776814	0.029936
NM_004587	RRBP1	-0.819226	0.036365
NM_002618	PEX13	-0.847441	0.016837
NM_004161	RAB1A	-0.869007	0.041363
NM_003099, 148955, 152826	SNX1	-0.897803	0.024230
NM_014426, 152227	SNX5	-1.002329	0.007902
NM_016530	RAB8B	-1.003628	0.021477
NM_007107	SSR3	-1.230001	0.022588
NM_022365	DNAJC1	-1.634016	0.000907

^a GenBank accession number

^b Signal Log ratio: power value is 2, for example 3.033860 is equal to $2^{3.033860}$.

^c For explanation of processing used to obtain FDR-adjusted *P*-value see materials and methods.

Appendix Table 3

Genes involved in Golgi apparatus altered in 435/BRMS cells relative to 435 cells.

Accession no. ^a	Symbol	Signal Log Ratio ^b	FDR adj P-Value ^c
Induced			
NM_012104, 138971-73	BACE1	2.427205	0.000663
NM_003762, 201994	VAMP4	1.090870	0.033237
NM_172341	PSENIEN	1.074803	0.013460
NM_016429	COPZ2	1.041162	0.027706
NM_006042	HS3ST3A1	1.016080	0.013820
NM_003916	AP1S2	0.997023	0.022876
NM_000585, 172174-75	IL15	0.966510	0.016706
NM_001003398, 001714	BICD1	0.916492	0.032477
NM_018714	COG1	0.774169	0.041457
NM_007190	SEC23IP	0.662779	0.041697
Repressed			
NM_003917, 080545	AP1G2	-0.593089	0.044733
NM_031231-32	APBA2BP	-0.624727	0.032046
NM_130468	D4ST1	-0.673337	0.039947
NM_002650, 058004	PIK4CA	-0.701046	0.039884
NM_003032, 173216-17	ST6GAL1	-0.867223	0.022725
NM_004161	RAB1A	-0.869007	0.041363
NM_003099, 148955, 152826	SNX1	-0.897803	0.024230
NM_000052	ATP7A	-0.924147	0.034246
NM_001497	B4GALT1	-0.988029	0.026170
NM_006577, 033252	B3GNT1	-1.029937	0.036966
NM_003574, 194434	VAPA	-1.108895	0.013738
NM_004737, 133642	LARGE	-1.144102	0.026161
NM_015003, 181076-77	GOLGA8A	-1.551530	0.039884
NM_006456	ST6GALNAC2	-1.565736	0.014191

^a GenBank accession number

^b Signal Log ratio: power value is 2, for example 2.427205 is equal to $2^{2.427205}$.

^c For explanation of processing used to obtain FDR-adjusted *P*-value see materials and methods.

Abbreviations

435-MDA-MB, 435 cells

ABI, Abl-interactor protein
 BP, biological process
 BRMS, breast cancer metastasis suppressor
 CC, cellular component
 DIAG, diacylglycerol
 Ece, endothelin converting enzyme
 ENPP, ectonucleotide pyrophosphatase phosphodiesterase
 FBS, fetal bovine serum
 GO, gene ontology
 HPRT, hypoxanthine guanine ribosyltransferase
 NK, natural killer cell
 PBS, phosphate buffered saline
 PKC, protein kinase C
 qRT-PCR, quantitative real time polymerase chain reaction
 RNA, ribonucleic acid
 USU, Utah State University
 435/BRMS1-MDA-MB, 435/BRMS1 cells
 ANOVA, analysis of variance
 BRCA, breast cancer gene
 CASP1, caspase 1
 DAG, directed acyclic graph
 DMEM, Dulbecco's modified eagle's medium
 EDTA, ethylene diamine tetraacetic acid
 ER, endoplasmic reticulum
 FDR, false discovery rate
 Her, human epidermal growth factor receptor
 MHC, major histocompatibility complex
 NF- κ -B, nuclear factor-kappa in B-cells
 PDGF, platelet derived growth factor
 pRb, retinoblastoma protein
 RMA, robust multi-array analysis
 Th, Helper T cells

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Top Biological Process GO Nodes

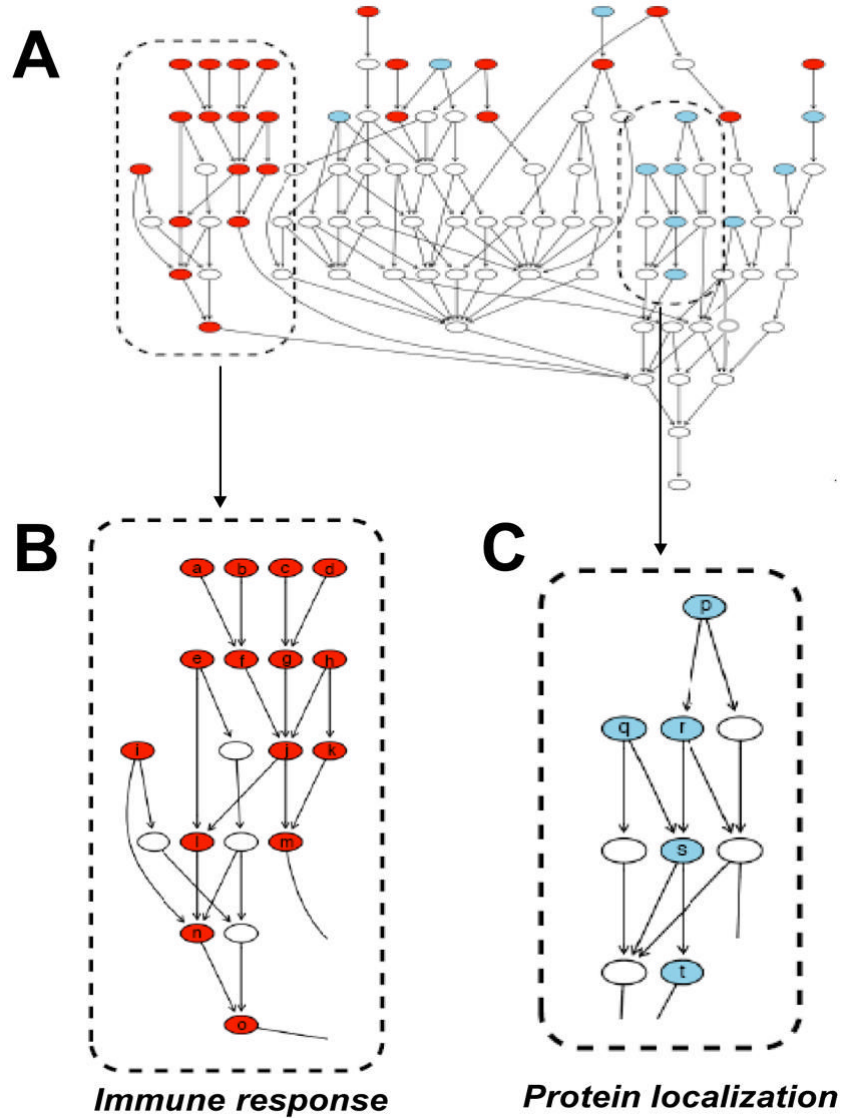


Figure 1. Directed acyclic graph of Gene Ontology (GO) terms over-represented by genes within the data set. Directed acyclic graph (DAG) of all biological process (BP) GO terms which met cut-off criteria (see materials and methods) with enlarged view of terms associated with immune response and protein localization. For all graphs red indicates BP GO terms that were generally induced in 435/BRMS1 cells while blue indicates BP GO terms generally repressed in 435/BRMS1 cells relative to 435 cells. Nodes represent general ancestor BP GO terms at the bottom of the graph proceeding upward to more specific child BP GO terms.

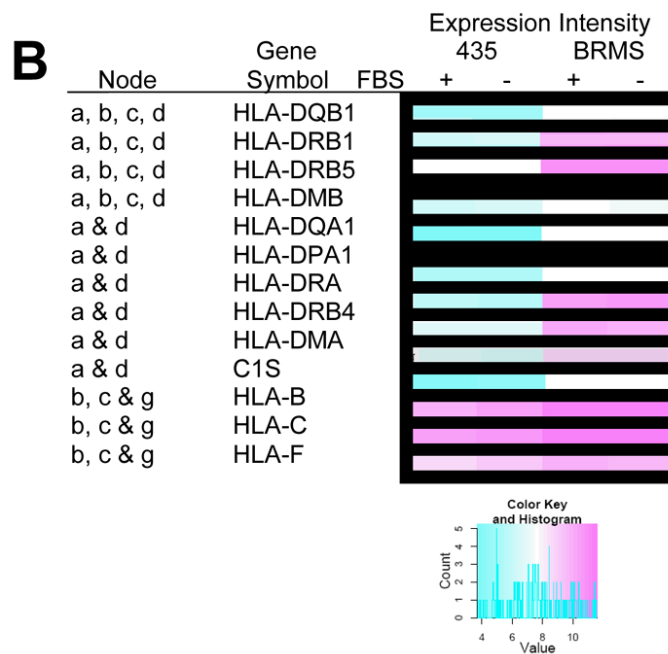
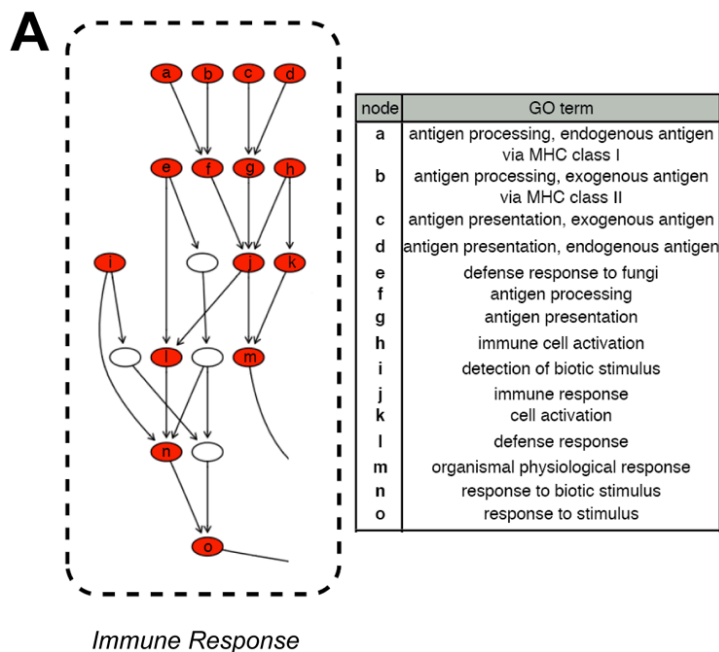


Figure 2. Immune response-associated Gene Ontology (GO) terms and related genes from the data set. A) GO terms are presented from the directed acyclic graph (DAG) of biological process (BP) GO terms which met the cut-off criteria (see materials and methods) with table identifying over-represented BP GO terms. Red nodes indicate BP GO terms which were generally induced in 435/BRMS1 cells relative to 435 cells. Nodes represent general ancestor BP GO terms at the bottom of the graph proceeding upward to more specific child BP GO terms. B) Several genes identified in some of the represented BP GO terms and their expression (as represented in a heat map), are shown in the table.

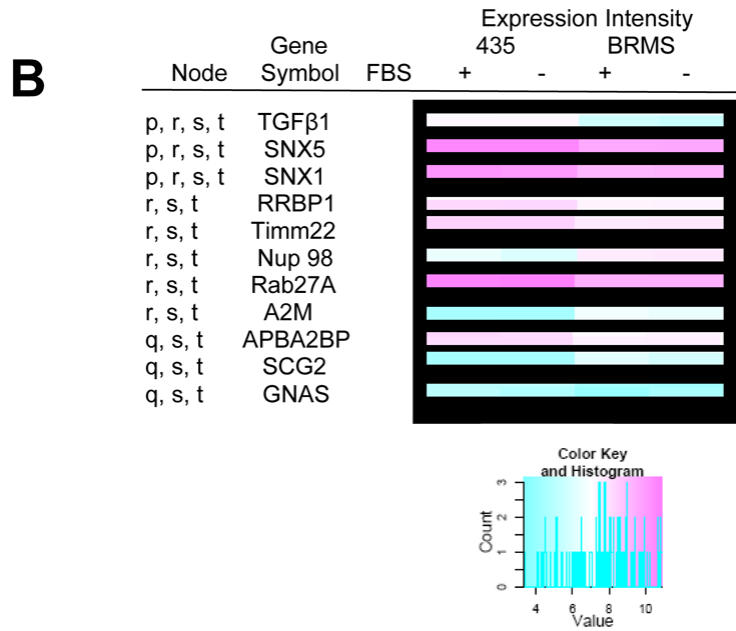
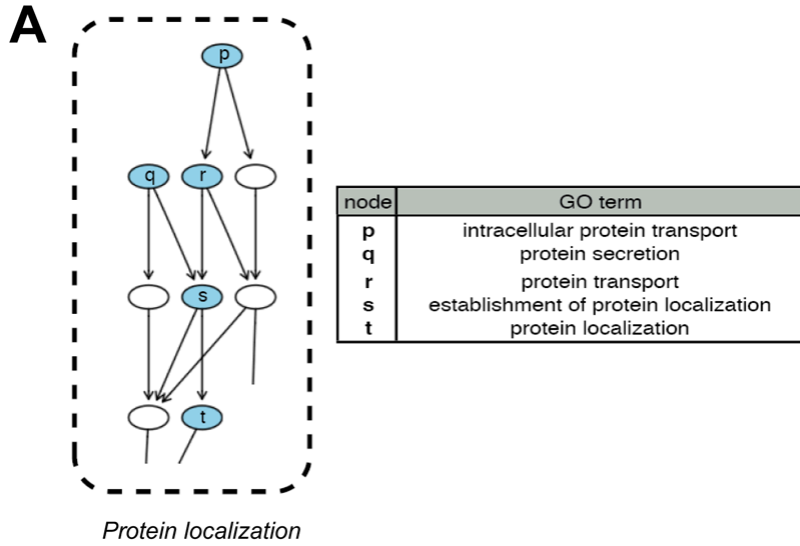


Figure 3. Protein localization-associated Gene Ontology (GO) terms and related genes from the data set. A) GO terms are presented from the directed acyclic graph (DAG) of biological process (BP) GO terms which met the cut-off criteria (see materials and methods) with table identifying over-represented BP GO terms. Blue nodes indicate BP GO terms which were generally repressed in 435/BRMS1 cells relative to 435 cells. Nodes represent general ancestor BP GO terms at the bottom of the graph proceeding upward to more specific child BP GO terms. B) Several genes identified in some of the represented BP GO terms and their expression (as represented in a heat map), are shown in the table.

Table 1

The top forty induced genes in 435/BRMS cells relative to 435 cells.

Accession no. ^a	Gene symbol	Function	Signal Log Ratio ^b	FDR adj. P-value ^c
NM_006169	NNMT	metabolism	6.163206	0.000737
NM_003785	PAGE1	immune response	5.383853	0.000035
NM_152997	C4orf7	unknown	5.344053	0.007440
NM_019111	HLA-DRA	immune response	5.268353	0.000562
NM_015444	RIS1	cell cycle, proliferation	4.663992	0.000663
NM_002122, 020056	HLA-DQA1/A2	immune response	4.639360	0.000035
NM_002124	HLA-DRB1	immune response	4.247076	0.000182
NM_000584	IL8	immune response	4.117897	0.002229
NM_002123	HLA-DQB1	immune response	4.110985	0.000035
NM_001056, 176825	SULT1C1	metabolism	4.110093	0.001317
NM_004114, 033642	FGF13	signal transduction	4.041216	0.000035
NM_032307	C9orf64	unknown	4.011214	0.000796
NM_001734, 201442	C1S	immune response	3.767461	0.003596
NM_181607	KRTAP19-1	cytoskeleton	3.721054	0.000803
NM_000200, 032866	HTN3/CGNL1	immune response	3.639734	0.000796
NM_002159	HTN1	immune response	3.432327	0.017445
NM_002124	HLA-DRB5	immune response	3.393892	0.000071
NM_021983	HLA-DRB4	immune response	3.367954	0.000075
NM_033554	HLA-DPA1	immune response	3.144453	0.000508
NM_004616	TSPAN8	signal transduction	3.139513	0.002499
NM_005708	GPC6	unknown	3.103186	0.001502
NM_004355	CD74	immune response, protein localization	3.033860	0.002536
NM_017863	CXorf48	DNA binding	2.988708	0.000326
NM_001452	FOXF2	transcription	2.956652	0.000237
NM_030812	LOC81569/ACTL8	cytoskeleton	2.920189	0.001461
NM_002571	PAEP	transport	2.886798	0.011394
NM_001187	BAGE	unknown	2.777546	0.016837
NM_001085	SERPINA3	immune response	2.550661	0.004990
NM_001445	FABP6	cell proliferation inhibitor, metabolism	2.477626	0.026161
NM_002727	PRG1	unknown	2.461429	0.001659
NM_012104, 138971-73	BACE1	protein modification, Golgi apparatus	2.427205	0.000663
NM_002121	HLA-DPB1	immune response	2.393904	0.001461
NM_004163	RAB27B	signal transduction	2.297115	0.038656
NM_001223, 033292-95	CASP1	signal transduction, apoptosis regulation	2.295373	0.017594
NM_021940	SMAP1	regulation of GTPase activity	2.281316	0.043648
-----	unknown	unknown	2.274630	0.030001
NM_001472, 001474-77, NM_021123	GAGE 2, 4-7B	cellular defense	2.259294	0.013588
NM_005195	CEBPD	transcriptional regulation	2.210539	0.003852
NM_003516	HIST2H2AA	unknown	2.175758	0.009671
NM_017855	APIN	unknown	2.173798	0.006827

^a GenBank accession number^b Signal Log ratio: power value is 2, for example 6.163206 is equal to 2^{6.163206}.^c For explanation of processing used to obtain FDR-adjusted P-value see materials and methods.

Table 2

The forty most repressed genes in 435/BRMS cells relative to 435 cells.

Accession no. ^a	Gene symbol	Function	Signal Log Ratio ^b	FDR adj. P-value ^c
NM_032525	TUBB6	unknown	-3.877230	0.005884
NM_000550	TYRP1	metabolism	-3.718408	0.003087
NM_000615, 181351	NCAM1	cell adhesion	-3.224920	0.000796
NM_014227	SLC5A4	transport - sugar and ions	-3.217907	0.010679
NM_005971, 021910	FXSD3	transport - ions	-3.019978	0.016583
NM_001647	APOD	transport - lipid	-2.853765	0.014191
NM_00100133, 016229	CYB5R2	transport - electron	-2.790324	0.001659
NM_006499, 201543-45	LGALS8	extracellular sugar binding	-2.782567	0.001461
NM_002615	SERPINF1	cell proliferation	-2.606028	0.005884
NM_00609	ATP8A1	transport - aminophospholipid	-2.563899	0.000818
NM_005944, 001004196	CD200	unknown	-2.436624	0.012907
NM_006877	GMPR	metabolism	-2.384433	0.000663
NM_017694	FLJ20160	unknown	-2.334477	0.012502
XM_372574	LOC390595	unknown	-2.315783	0.017422
NM_007078	LDB3	cytoskeleton, protein binding	-2.312725	0.016837
NM_000212	ITGB3	cell adhesion, signal transduction	-2.248644	0.007499
-----	LOC401022	unknown	-2.228267	0.002154
NM_005761	PLXNC1	cell adhesion	-2.226256	0.036549
NM_005328	HAS2	ECM component metabolism	-2.152467	0.015516
NM_152309	PIK3AP1	lipid binding	-2.150887	0.005745
-----	LOC283824	unknown	-2.124596	0.013738
NM_024505	NOX5	cell proliferation, cytokine secretion	-2.094189	0.030001
NM_007085	FSTL1	extracellular heparin binding	-2.086846	0.001659
NM_053042	KIAA1729	nucleic acid binding	-2.073675	0.013882
NM_152270	FLJ34922	unknown	-1.977802	0.026823
NM_203463	LASS6	cell proliferation, metabolism	-1.948670	0.001461
NM_014654	SDC3	cytoskeletal protein binding	-1.928120	0.016836
NM_001002034	LOC150368	unknown	-1.914849	0.000631
NM_012228	MSRB2	transcription factor	-1.907740	0.019284
NM_014782, 177949	ARMCX2	binding	-1.905816	0.010707
-----	KIAA1295	unknown	-1.900952	0.007121
NM_000070, 024344, 173087-90, 212464-65, 212467	CAPN3	signal transduction, cell adhesion	-1.866039	0.026170
NM_006236	POU3F3	transcription factor	-1.829551	0.002536
NM_005380, 182744	NBL1	apoptosis	-1.829396	0.000799
NM_000935, 182943	PLOD2	protein modification	-1.828234	0.001461
NM_003500	ACOX2	metabolism	-1.793333	0.031810
NM_015150	RAFTLIN	unknown	-1.776159	0.015190
NM_005160	ADRBK2	signal transduction	-1.763304	0.015429
NM_003812	ADAM23	cell adhesion	-1.751657	0.001461
NM_000660	TGFB1	cell proliferation, signal transduction	-1.751519	0.013738

^a GenBank accession number

^b Signal Log ratio: power value is 2, for example -3.877230 is equal to $2^{-3.877230}$.

^c For explanation of processing used to obtain FDR-adjusted P-value see materials and d. methods.