

A method for visualization of “omic” datasets for sphingolipid metabolism to predict potentially interesting differences[§]

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Abstract Sphingolipids are structurally diverse and their metabolic pathways highly complex, which makes it difficult to follow all of the subspecies in a biological system, even using “lipidomic” approaches. This report describes a method to use transcriptomic data to visualize and predict potential differences in sphingolipid composition, and it illustrates its use with published data for cancer cell lines and tumors. In addition, several novel sphingolipids that were predicted to differ between MDA-MB-231 and MCF7 cells based on published microarray data for these breast cancer cell lines were confirmed by mass spectrometry. For the data that we were able to find for these comparisons, there was a significant match between the gene expression data and sphingolipid composition ($P < 0.001$ by Fisher's exact test). Upon considering the large number of gene expression datasets produced in recent years, this simple integration of two types of “omic” technologies (“transcriptomics” to direct “sphingolipidomics”) might facilitate the discovery of useful relationships between sphingolipid metabolism and disease, such as the identification of new biomarkers.—Momin, A. A., H. Park, B. J. Portz, C. A. Haynes, R. L. Shaner, S. L. Kelly, I. K. Jordan, and A. H. Merrill, Jr. **A method for visualization of “omic” datasets for sphingolipid metabolism to predict potentially interesting differences.** *J. Lipid Res.* 2011. 52: 1073–1083.

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The importance of sphingolipids for cell structure and regulation (1–3) has inspired the development of mass spectrometric methods to analyze key categories of these compounds (see Refs. 4–8 for examples). These cover hundreds of subspecies but still miss a large portion of the

“sphingolipidome” due to its size and complexity (including the existence of many isomers and isobars) (9) and wide range in amounts. Therefore, it would be useful to have tools that draw on other sources of information, such as gene expression data, to direct attention toward particular subspecies that might otherwise be overlooked by mass spectrometry or other methods.

This article describes a pathway visualization tool to predict differences in sphingolipid composition using microarray data, then explores its utility using publicly available datasets for cancer cells in culture and tumors. Cancer was selected because sphingolipids are involved in many processes that are important in regulation of growth, survival, adhesion, and migration (10), and tumors often display alterations in sphingolipid composition (11, 12), some of which have been proposed as cancer markers (13–15). Furthermore, there have been a large number of

Abbreviations: 3KSA, 3-ketosphinganine; BCA, bicinchoninic acid; Cer, ceramide; CerP, ceramide 1-phosphate; CerS, ceramide synthase (isoforms 1–6); CERT, Cer transporter; DEGS, dihydroceramide desaturase; DHCer, dihydroceramide; DH, dihydro- backbone; ELOVL, fatty acyl-CoA elongase; EP, ethanolamine phosphate; Fut, fucosyltransferase; Fut1/Fut2, alpha(1,2)fucosyltransferase; GalCer, galactosylCer; Gb3 or CTH, ceramide trihexoside; Gb4, ceramide tetrahexoside (globoside); GlcCer, glucosylCer; GSL, glycosphingolipid; HexCer, monohexosylCer; HNSCC, head and neck small cell carcinoma; IDC, invasive ductal carcinoma; KEGG, Kyoto Encyclopedia of Genes and Genomes; LacCer, lactosylceramide; LCB, long chain base; phyto-Cer, phytoceramide; phyto-HexCer, monohexosyl phytoceramide; phyto-SM, phytosphingomyelin; Sa, sphinganine; Sa1P, sphinganine 1-phosphate; SCD, stearyl-CoA desaturase; SL, sphingolipid; SM, sphingomyelin; SPTLC1, serine palmitoyltransferase 1; SPTLC2, serine palmitoyltransferase 2; SPTLC3, serine palmitoyltransferase 3; SPHK, sphingosine kinase (isoforms 1 and 2); So, sphingosine; So1P, sphingosine 1-phosphate; SSEA-3, stage-specific embryonic antigen 3; SSEA-4, stage-specific embryonic antigen 4; ST, sulfatide; UGCG, UDP-glucose ceramide glucosyltransferase; UGT1, glucosylceramide synthase; UGT8, galactosylceramide synthase.

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[§]The online version of this article (available at <http://www.jlr.org>) contains supplementary data in the form of *i*) one table with accompanying text, *ii*) online tutorial screens, *iii*) two figures, and *iv*) 59 heat maps.

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microarray analyses of gene expression in cancer cell lines and tumors, and more are certain to be conducted; therefore, a tool that can utilize this data to direct attention to compounds that are potentially different in cancer might aid the understanding of the molecular changes associated with this disease and possibly uncover novel biomarkers.

There are ample precedents for this approach. For example, elevated expression of mRNA for GD3 synthase has been correlated with the amounts of GD3 in melanoma cell lines (16), and in the reverse direction, the finding of low amounts of C18-ceramide (Cer) in head and neck cancer led to the discovery of lower expression of CerS1 (17). Nonetheless, there is currently no accurate pathway map for the visualization of datasets for large categories of genes and metabolites involved in sphingolipid biosynthesis. Therefore, the goal of these studies was to develop tools for this purpose by using PathVisio v2 (18), an open access pathway browser, and updated and expanded maps to visualize data from analyses of gene expression and sphingolipid metabolites. A preliminary version of this tool was described earlier (19); this article provides more information and illustrates its application to cancer cell lines, tumors, and normal tissues, which found substantial agreement between gene expression and metabolite profiles.

MATERIALS AND METHODS

Materials

MCF7 cells were purchased from the ATCC (Manassas, VA), and MDA-MB-231 cells were provided by Dr. Nathan Bowen (School of Biology, Georgia Institute of Technology, Atlanta, GA). Cell culture medium was procured from Gibco (Carlsbad, CA), and FBS was obtained from Hyclone (Logan, UT). U-¹³C labeled palmitic acid was obtained from Cambridge Isotopes (Andover, MA). The mass spectrometry internal standards were procured from Avanti Polar Lipids (Alabaster, AL); other glycosphingolipid (GSL) standards were obtained from Sigma (St. Louis, MO) and Matreya (Pleasant Gap, PA). All other reagents and solvents were HPLC grade.

Modification of PathVisio v2 and construction of sphingolipid pathway maps

PathVisio v2 (18), open source Java software, was selected for the construction of pathway maps for different branches of sphingolipid biosynthesis and the visualization of differentially expressed genes. A tutorial on how to use Path Visio v2 and the tools described in this manuscript is available on the web site <http://sphingolab.biology.gatech.edu/sphingoPathvisioTutorial1.html>. To differentiate metabolites from genes, the former were displayed as circular objects, while the latter had a rectangular shape. New pathway maps using PathVisio v2 were constructed by altering Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway diagrams to include metabolite identifiers at the nodes, while the genes were updated to reflect new discoveries and changes in nomenclature. Some of the prominent changes in the sphingolipid pathway maps are as follows: the addition of ceramide synthase gene isoforms (CerS1-6) (20–24), dihydroceramide desaturase (DEGS1, DEGS2) (25), and α (1,2)fucosyltransferases (FUT1 and FUT2) (14). The layout of the backbone biosynthesis map was published previously (19) and is provided

in supplementary Fig. 1; it was modified to include branches for the synthesis of 3-ketodihydroceramides (26), sphingolipids with dihydroceramide backbones, and phyto-sphingolipids. These modified KEGG pathway maps, as well as ones for gangliosides and globo series GSL, can be obtained from www.sphingomaps.org or the Wikipathways database (27). A list of all the genes is provided in **Table 1** and supplementary online data.

Selection of DNA microarray datasets and statistical analysis

Keyword searches were conducted on National Center for Biotechnology Information, Gene expression omnibus (NCBI GEO) (28) Array express (29) and Oncomine (30) for “cancer type” or “cancer cell line,” and the outcomes were examined to identify either expression datasets having multiple replicates of control and cancer tissues or all the cell lines to be compared. The Affymetrix expression files (.CEL file) for the selected datasets were downloaded and processed by Affymetrix expression console v1.1 (Affymetrix, Santa Clara, CA) using the MAS5 algorithm. For non-Affymetrix studies, the normalized and background-corrected expression datasets (GDS file, NCBI GEO format) were accessed using the GEOquery package (31) of Bioconductor v1.8 (32) and exported as tab-delimited text files. Sphingolipid-specific gene expression values were filtered using a Perl search algorithm and a list of preselected Affymetrix probe IDs or gene ID (see below). The gene expression fold change was determined using Excel (Microsoft, Seattle, WA), after which the dataset was formatted as a PathVisio input file (comma-separated values) with “gene ids,” “system code,” and “fold change.”

Selection of gene probes

To identify the best probes for sphingolipid genes, the sequences of all probes directed to individual transcripts were retrieved from the Affymetrix Netaffx tool (33) and compared with the mRNA using Blat search (34) in the University of California, Santa Cruz (UCSC) genome browser (35). The probe that recognized the catalytic transcript and hybridized nearest to its 3' UTR was selected to represent the gene for pathway model visualization, and others were excluded. The procedure was repeated for all sphingolipid genes for Affymetrix chips HG-U95 A-E and HG-U133 plus 2 (supplementary online data).

Perl algorithm for extraction of expression values

A Perl script was written to search for and extract gene expression values for sphingolipid-related transcripts. The program transforms the list of probe IDs or gene IDs into an array, and then searches for similar objects in a normalized gene expression data file, where the rows represent probe IDs and columns individual experiments. The loop function efficiently extracts and writes the sphingolipid gene expression values into a text file, which is used to calculate gene fold changes. A copy of the Perl script and instructions for its use are provided in the supplementary online data.

Pathway visualization

In the PathVisio v2 browser, the appropriate gene and metabolite database was selected, and the expression dataset was created after importing the saved input file using the Import Expression Dataset tab. The color criteria for the expression dataset were specified using the Visualization Option function. Gene and metabolite expression changes were visualized on the modified sphingolipid pathway maps by opening the pathway diagram in PathVisio v2, and then selecting the saved expression dataset and visualization criteria. The pictures were prepared by exporting the images as PDF files; while changes to font colors were performed

TABLE 1. Genes involved in de novo sphingolipid biosynthesis in Fig. 1

Gene ID	Gene Name	Ref.
SPTLC1, SPTLC2, SPTLC 3	Serine palmitoyltransferase subunits (1, 2, and 3)	(52)
KDSR, 3KSR, FVT1	3-Ketosphinganine reductase	(53)
CerS1, CerS2, CerS3, CerS4	Ceramide synthase (1-6)	(20-24)
CerS5, CerS6		
SPHK1, SPHK2	Sphingosine kinase (1, 2)	(54)
SGPL1	Sphingosine 1-phosphate lyase	(55)
DES1, DES2, DEGS1, 2	Dihydroceramide desaturase (1 and 2)	(25)
ASAH1, ASAH2, ASAH3, ASAH3L, PCHA	Ceramidases (Acid, neutral, phytoceramide)	(56-60)
SMS1, SMS2, SMSr,	Sphingomyelin synthase (1, 2, and atypical)	(40, 61)
UGCG, Gba	Glucosylceramide synthase	(62)
UGT8, CGT	Galactosylceramide synthase	(63)
CERK	Ceramide kinase	(64, 65)
CERT	Ceramide transporter	(42, 66)
SMPD1, SMPD2, SMPD3, SMPD4	Sphingomyelinases (acidic, neutral)	(67, 68)
B4GALT6	Lactosylceramide synthase	(69)
GAL3ST1	galactose-3-O-sulfotransferase 1	(70)
ELOVL 1-7	Fatty acid CoA elongases (1-7)	(71)
SCD	Stearoyl CoA desaturase	(72)
ssSPTa, ssSPTb	Small subunits of SPT	(73)
ORMDL1, ORMDL 2, ORMDL 3	Orosomucoid (1-3)	(74, 75)

The table lists alternative gene abbreviations and complete names for genes encoding enzymes for different steps of de novo sphingolipid biosynthesis displayed in Fig. 1. Genes involved in the biosynthesis of glycosphingolipids are described in a table in the supplementary online data.

with CorelDraw X4 (Mountain View, CA) as summarized in the supplementary online data. A detailed tutorial for this tool (including how to download PathVisio software) is available (see the supplementary online data).

Estimation of the relative amounts of the sphingolipids for the samples used in the gene expression comparisons

The MCF7 and MDA-MB-231 cells were grown in 100 mm tissue culture dishes using RPMI 1640 medium containing 5% FBS under conditions described previously (36). When the cells reached 80-90% confluency, they were washed once with PBS solution, before being scraped into screw capped glass test tubes for sphingolipid extraction as previously described (37). An aliquot of the cells was retained from each dish to measure the protein concentration by the bicinchoninic acid (BCA) assay (Pierce Biotechnology, Rockford, IL). During stable isotope-labeling studies, the media was supplemented with 0.1 mM [^{13}C]palmitic acid complexed with equimolar fatty acid free BSA (Calbiochem, Gibbstown, NJ). After the times described for each experiment, the sphingolipids were analyzed by LC ESI-MS/MS using ABI 3000 and ABI 4000 LC-tandem mass spectrometers (Applied Biosystems, Foster City, CA), as previously described (37).

When available, published quantitative data for metabolites were used to compute fold changes. In some instances, only photographs of TLC plates were provided, and in those cases, crude estimates were obtained from the relative densities obtained using Fluorchem 5500 (Alpha Innotech, San Leandro, CA). The sources of these data are provided in the text.

RESULTS AND DISCUSSION

Summary of the sphingolipid biosynthetic pathway as the basis for construction of maps to visualize gene expression and metabolite data in a pathway context

Fig. 1 displays the pathway map for the initial steps of sphingolipid biosynthesis that we constructed for use with PathVisio v2 (18). Metabolites are in circles, and the genes associated with each step of the pathway are in boxes. The

heatmap coloring of Fig. 1 is a comparison of MDA-MB-231 and MCF7 cells, which will be discussed later in this article (Figs. 2–5). The gene annotations for the individual steps of the pathway are listed in Table 1, and additional details are given in supplementary Fig. 1 and recent reviews (9, 19). An expanded form of the pathway has been provided to www.wikipathways.com (27) and can be downloaded from that site.

As displayed, de novo sphingolipid biosynthesis is initiated in the upper-left corner (Fig. 1, dashed box A) by the condensation of serine and palmitoyl-CoA by serine palmitoyltransferase (SPT) to produce 3-ketosphinganine (3K_{Sa}). The five genes for SPT represent the three genes known to code for functional proteins (SPT1-3 from SPTLC 1, 2, and 3) and two shorter transcripts (ssSPTa and ssSPTb) that have been suggested to regulate SPT (38). The 3K_{Sa} is reduced to sphinganine (Sa), which is at a branch point where it can be acylated by ceramide synthases (CerS, in dashed box B²) or phosphorylated by sphingosine kinase (SPHK1 or 2) to Sa 1-phosphate (SaIP), which can be terminally degraded to hexadrenal (C16:0-al) and ethanolamine phosphate (EP). The acylation products, dihydroCer (DHCer), are at another branch point (dashed box C) and are desaturated to Cer by DES1 (and possibly DES2), hydroxylated to 4-hydroxyDHCer (phyto-Cer) by DES2 (25), or converted to a more complex dihydrosphingolipids, as will be discussed later. DHCer can also be hydrolyzed by ceramidases (ASAH 1-3, 3L, and PCHA). For most cells, the desaturation branch is presumed to predominate because Cer is the major lipid backbone of most complex sphingolipids (9, 39).

²The fatty acyl-CoA elongases and desaturases that biosynthesize the cosubstrates for CerS are also shown in this diagram and in supplementary Fig. 1.

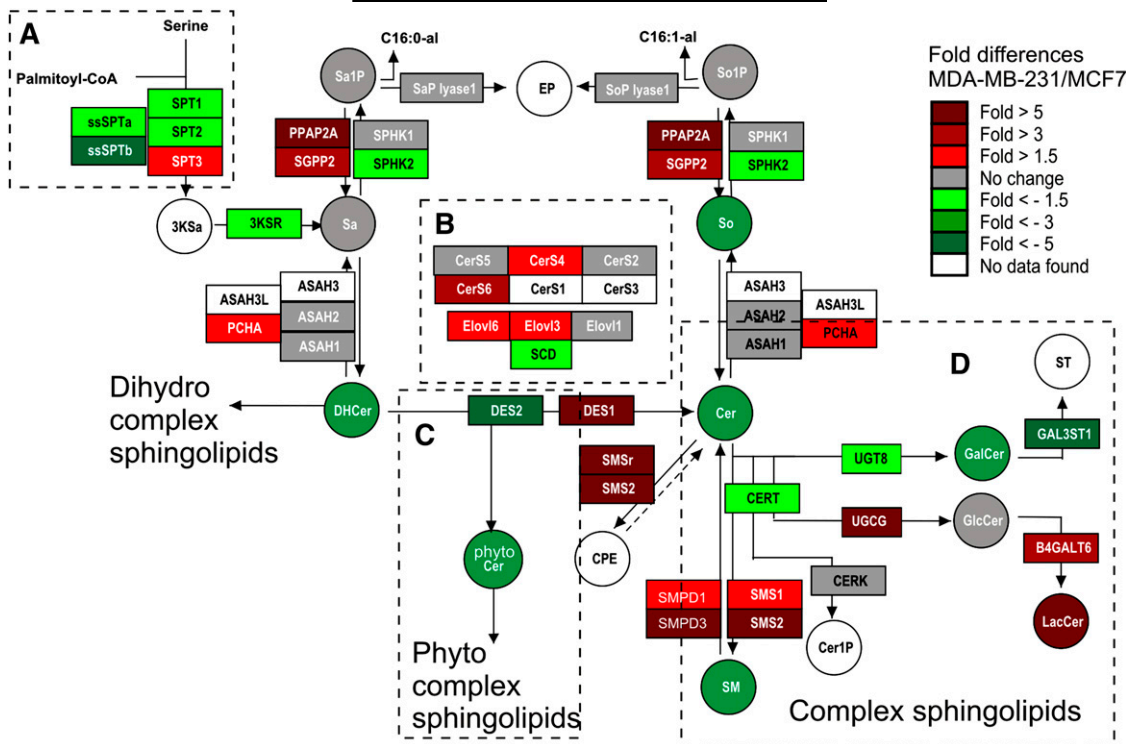


Fig. 1. Visualization of transcription and metabolite differences in early steps of sphingolipids biosynthesis between MDA-MB-231 and MCF7 cells using PathVisio v2 with an expanded pathway map. The figure depicts sphingolipid genes (rectangles) and metabolites (circles) that participate in early steps of sphingolipid biosynthesis prepared using PathVisio v2 (for more detail, see supplementary Fig. 1) (18). The updated and expanded pathway map was overlaid by the gene expression fold change difference between MDA-MB-231 and MCF7 cells obtained from a previous study (36). Variations in sphingolipid composition (shaded circles) between the two cell lines were determined by LC-ESI-MS/MS as previously described (4) and compared with gene expression as highlighted by circumscribed dash boxes (A-D) and discussed in Figs. 3–6. The box and circle shades represent the degree of upregulation and downregulation indicated by the color chart.

Cer is also at a branch point (illustrated in dashed box D for Cer) for the biosynthesis of sphingomyelin (SM) (40) (and a recently reported Cer-phosphoethanolamine) (41), glucosylCer (GlcCer), galactosylCer (GalCer), or Cer 1-phosphate (Cer1P). Shown in this scheme is CERT, an additional protein that transports Cer to the trans-Golgi for biosynthesis of SM, CerP, and possibly some GlcCer (42). These reactions sometimes take place with DHCer and phyto-Cer, but for simplicity, these additional steps have not been included in Fig. 1. Pathway maps can be customized for PathVisio v2 at the discretion of the user, which allows inclusion of additional steps, such as genes coding for the enzymes of sphingolipid turnover (see Fig. 1 for the sphingomyelinases but not the glycosidases for GlcCer and GalCer), and for updating as new genes and metabolites are discovered.

This particular scheme displays the next metabolic step for GlcCer, glycosylation of GlcCer to lactosylCer (LacCer), after which the pathway branches into five so-called root structure series (globo-series, isoglobo-series, ganglio-series, lacto-series and neo-lacto-series) (9), and for GalCer, which is primarily converted to sulfatides (ST). Separate pathway maps are shown for two subcategories of more complex glycosphingolipids, gangliosides (Fig. 6), and globo-series neutral glycosphingolipids (Fig. 7) (these have also been made available through Wikipathways) (27). Background references for the genes

for these steps are given in the supplementary online data. As noted before, users of PathVisio v2 can update and expand these pathways as new information is obtained.

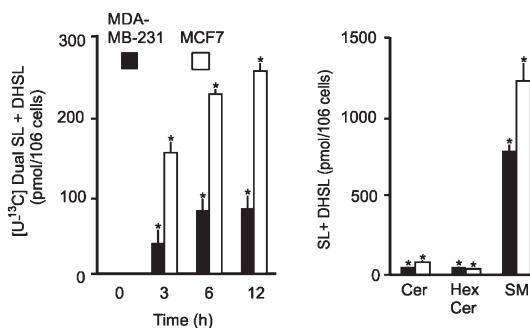


Fig. 2. Comparison of de novo sphingolipid biosynthesis between MDA-MB-231 and MCF7 cells with gene expression of serine palmitoyltransferase. The two cell lines were grown as previously described (36) and incubated with and without 0.1 mM [¹⁴C]palmitic acid as the BSA complex for 0, 3, and 6 h. A: The sphingolipids labeled in sphingoid base and fatty acid (Dual) were measured by LC-ESI-MS/MS (4). B: In parallel experiments, unlabeled sphingolipids were also measured for both cell types. A: Sum of Dual [¹⁴C]palmitate-labeled complex sphingolipids with SL (Cer) and DHSL (DHCer) backbone. B: Sum of unlabeled complex SL and DHSL in ceramide (Cer), monohexosylceramide (HexCer), and sphingomyelin (SM). The significance, indicated by an asterisk, was determined by a *t*-test at *P* < 0.05 for *n* = 3 (labeling study) and *n* = 6 (for unlabeled cells).

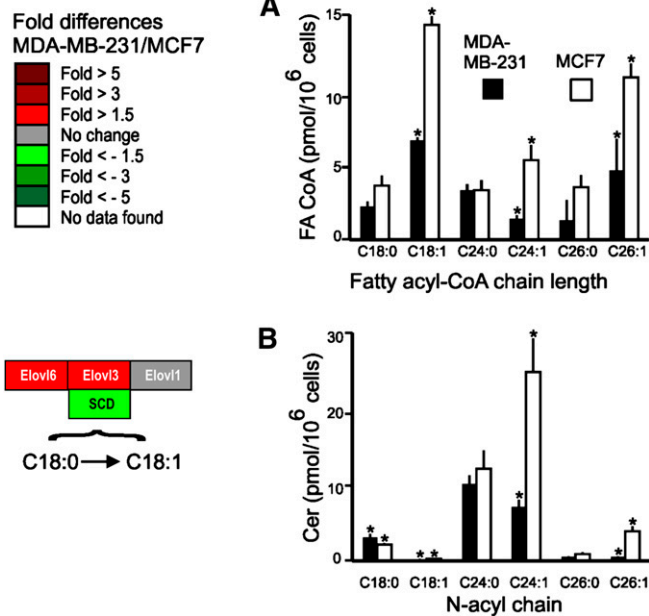


Fig. 3. Comparison of the composition of fatty acyl CoA and ceramide chain lengths for MDA-MB-231 and MCF7 cells. MDA-MB-231 and MCF7 cells were grown as previously described (36). The fatty acyl-CoA (FA-CoA) and ceramides (Cer) were measured by LC-ESI-MS/MS (37, 84). **A:** Composition of saturated and monounsaturated FA-CoA chain lengths. **B:** N-acyl chain distribution of Cer corresponding to the FA-CoA measured in **A**. The significance, indicated by an asterisk, was determined by a *t*-test at $P < 0.05$ for $n = 6$.

Preparation of data for visualization in sphingolipid gene expression pathway diagrams

Before discussing the heatmap differences in these figures, some comments will be made about the datasets and the formatting of the data for visualization using PathVisio v2. Cancer was selected as the source of the datasets because published gene expression data are available for cancer cells, normal cells, and tumor cells (36, 43, 44) from public

microarray repositories (28, 29). Platform-specific tools were used to normalize raw expression values for samples within studies as described under “Methods,” and a Perl script was written (see supplementary online data) to sort out expression values for transcripts involved in sphingolipid biosynthesis. Fold differences were computed either by taking the ratio between two conditions (such as two cell lines or cancer stages) or by normalizing the individual gene expression by the average among all samples (in the case of multiple cell lines), and then visualized using the sphingolipid biosynthesis pathway maps and PathVisio v2.

The presence of multiple probes for individual transcripts and a large number of transcripts posed a challenge during the analysis. To address the former, probes specific to the 3′ UTR of the mRNA were chosen by a Blat search (34) using the UCSC genome browser (35). A list of the optimum probes for sphingolipid biosynthesis in human HG-U133 Plus2 and HG-U95 Affymetrix gene chips is provided in the supplementary online data. In case of the latter, a search algorithm was written in Perl (supplementary online data) to sort out the expression values for the optimum probes using the list of preselected probes.

Rules for correlating mRNA and metabolite abundance

To study the correlation between transcript and metabolite amounts, certain rules were followed based on biochemical understanding: *a)* Only the de novo biosynthesis steps beginning with the condensation of serine and palmitoyl-CoA were evaluated; the reverse steps were usually not included; *b)* For isoforms that are known to display distinct specificities, such as SPTLC3, the CerS isoforms (CerS1-6), ELOVL 1-6, SCD, and DES2 (versus DES1), the changes in individual transcripts were compared with their corresponding sphingolipid subspecies; *c)* Enzymatic steps represented by multiple gene isoforms without known differences in the specificities of their product enzyme were compared with the total metabolite amount (in addition, the step was

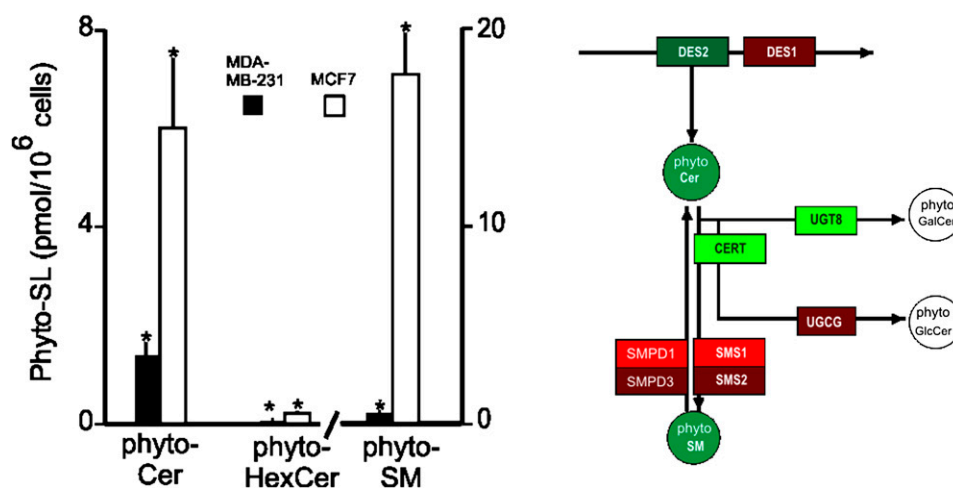


Fig. 4. Comparison of the phyto-sphingolipids of MDA-MB-231 and MCF7 cells. LC-ESI-MS/MS comparison of phyto-sphingolipids (phyto-SL; i.e., with a 4-hydroxysphinganine backbone), phyto-ceramide (phyto-Cer), phyto-mono-hexosylceramide (phyto-HexCer), and phyto-sphingomyelin (phyto-SM) (note different scale) for MDA-MB-231 and MCF7 cells. The significance, indicated by an asterisk, was determined by a *t*-test at $P < 0.05$ for $n = 6$.

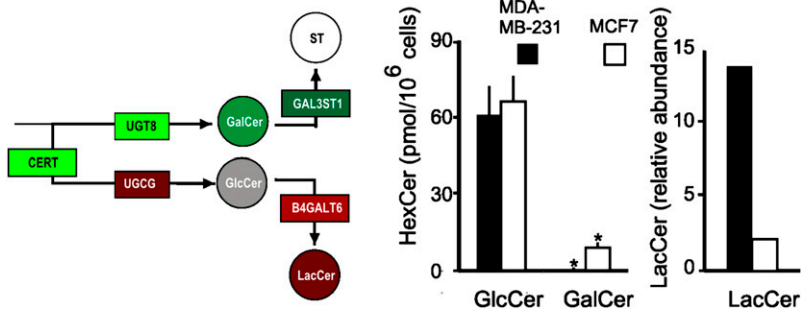


Fig. 5. Comparison of monohexosylceramides and lactosylceramide of MDA-MB-231 and MCF7 cells. Results from LC-ESI-MS/MS analysis of GlcCer and GalCer of MDA-MB-231 and MCF7 cells as previously described (4). The significance of differences, indicated by an asterisk, was determined by a *t*-test at $P < 0.05$ for $n = 6$. LacCer was estimated by densitometric analysis of LacCer data from a previous study (48).

considered “upregulated” if at least one gene isoform was elevated, the step was considered “downregulated” if all isoforms were downregulated, or the step was considered “unchanged” if all isoforms were unchanged or there was a mix of unchanged and reduced); and *d*) correlations were only counted when there was data for both the relative gene expression and accompanying metabolites, which is obviously a small subset of all of the potential molecules.

To test the utility of this technique, gene expression datasets from breast cancer cell lines were compared first, then normal breast cells and tumor cells were examined as described in the following sections.

Comparison of gene expression and sphingolipid differences between MCF7 and MDA-MB-231 breast cancer cells

Examination of the apparent³ gene expression differences between the highly metastatic ER-negative MDA-MB-231 and ER-positive MCF7 cell lines (Fig. 1) leads to a

number of relatively self-evident predictions, including that MDA-MB-231 cells might have: *a*) lower de novo sphingolipid biosynthesis due to lower mRNA abundance of SPTLC1 and SPTLC2 (Fig. 1, dashed box A), whereas higher expression of SPTLC3 might indicate an elevation of sphingolipids with a 16-carbon sphingoid base backbone; *b*) lower proportions of Cer with unsaturated fatty acyl chains (e.g., C24:1 and C26:1) due to lower SCD, the enzyme that converts C18:0-CoA to C18:1-CoA, which is subsequently elongated to C24:1- and C26:1-CoA (45); *c*) lower phyto-sphingolipids in MDA-MB-231 cells due to lower DES2 (Fig. 1, dashed box C); and *d*) lower GalCer production but higher GlcCer and/or LacCer due to lower UGT8 and higher UDP-glucose ceramide glucosyltransferase (UGCG) and B4GALT6 (Fig. 1, dashed box D). In the interest of space, the complete figures comparisons of mRNA abundance between the two cells for other branches of de novo sphingolipid and glycosphingolipid biosynthesis are displayed in supplementary Fig. II.

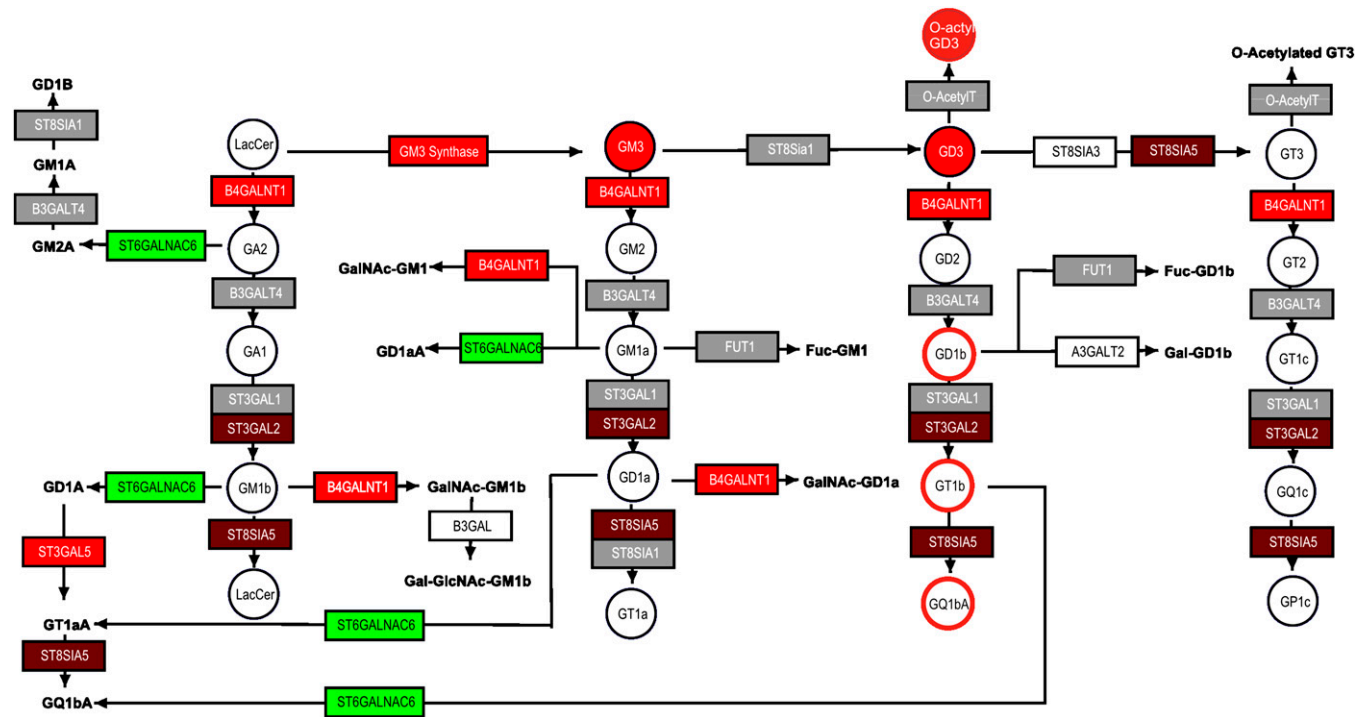


Fig. 6. Comparison of microarray data and metabolites for ganglioside biosynthesis for invasive ductal breast carcinoma and normal ductal tissue. Gene expression differences between the two breast cancer stages were visualized using data from a previous study (43) and updated ganglioside biosynthetic maps in PathVisio v2 (18). Ganglioside measurements are from reports using similar tumors (circles with a solid shade) (49) and patient serum (circles with colored edges) (15). The color scale for the difference between MDA-MB-231/MCF7 cells is the same as in Fig. 1.

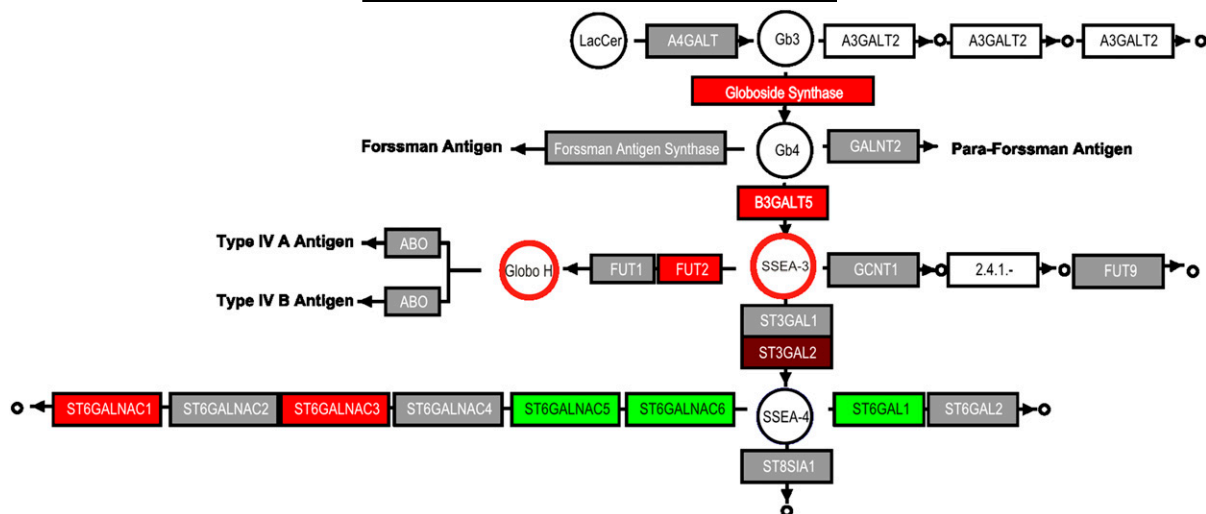


Fig. 7. Comparison of microarray data and corresponding metabolites for globo-series glycosphingolipid biosynthesis for invasive ductal breast carcinoma and normal ductal tissue. Comparison of gene expression differences between normal and invasive ductal carcinoma (IDC) from a previous study (43) using updated KEGG pathway maps for globo-series biosynthesis to metabolite changes previously reported using flow cytometry (14). The color scale for the difference between MDA-MB-231/MCF7 cells is the same as in Fig. 1.

To test the prediction that MDA-MB-231 cells have lower de novo sphingolipid biosynthesis, cells were incubated with [$U\text{-}^{13}\text{C}$]palmitic acid, and the backbone and fatty acid-labeled sphingolipids (i.e., unambiguously biosynthesized de novo) were quantified by LC ESI-MS/MS. At each time point, there was less incorporation of ^{13}C -palmitate into the sphingolipids of MDA-MB-231 cells than in those of MCF7 cells. Differences in the composition of these cell lines was also seen in LC ESI-MS/MS analysis of the unlabeled cells (shown in Fig. 2B), with MDA-MB-231 cells having 48.8 ± 4.1 Cer/ 10^6 cells and 783 ± 40 pmol SM/ 10^6 cells (Fig. 2B, black bars), which was significantly lower than in MCF7 cells, which had 79.8 ± 11.1 pmol Cer/ 10^6 cells ($P < 0.001$) and $1,220 \pm 110$ pmol SM/ 10^6 cells ($P < 0.001$) (Fig. 2B, white bars). In contrast, the amounts of unlabeled HexCer were higher in MDA-MB-231 cells, 50.3 ± 3.4 pmol HexCer/ 10^6 cells, than in MCF7 cells, 36.9 ± 5.9 pmol HexCer/ 10^6 cells ($P < 0.001$). Overall, however, both the ^{13}C -labeling and the SL amounts in MDA-MB-231 cells were lower than those in MCF7 cells, which were consistent with the apparent differences in SPT1 and SPT2 (Fig. 1).

Another apparent difference between MDA-MB-231 and MCF7 cells is a higher expression of SPTLC3, which has been proposed to metabolize a shorter chain (d16:1) (46, 47). Although standards are not yet available for a rigorous quantitation of d16:1-containing sphingolipids, preliminary analysis of the Cer subspecies of these cells by LC ESI-MS/MS (4) indicated that MDA-MB-231 cells had a higher proportion of d16:1-Cer (1.9 ± 0.4 pmol/ 10^6 cells) than did MCF7 cells (1.1 ± 0.2 pmol/ 10^6 cells) ($P < 0.05$,

³These differences are referred to as “apparent” because microarray data are not quantitative and should be validated by another method before drawing mechanistic conclusions. However, the purpose of this study was not to establish the validity of the gene expression data per se, but rather to explore whether publicly available microarray data can aid sphingolipid analysis by directing attention to possible differences.

$n = 3$), which is agreement with the apparently higher SPTLC3 expression in MDA-MB-231 cells (Fig. 1).

To test the prediction from the microarray data that MDA-MB-231 cells would have lower amounts of SL with unsaturated fatty acids (due to production of less mono-unsaturated fatty acyl-CoAs because SCD expression is lower) (Fig. 1, dashed box B), the FA-CoAs and the chain-length distribution of the Cer were determined for the two cell lines. As shown in Fig. 3, MDA-MB-231 cells had significantly lower amounts of all of the major mono-unsaturated fatty acyl-CoAs (i.e., per 10^6 MDA-MB-231 cells: 6.3 ± 0.22 pmol of C18:1, 1.4 ± 0.2 pmol of C24:1, and 4.6 ± 2.1 pmol C26:1; compared with MCF7 cells: 14.3 ± 0.6 pmol C18:1/ 10^6 cells, $P < 0.001$; 5.3 ± 1 pmol C24:1/ 10^6 , $P < 0.001$; and 10.8 ± 0.9 pmol C26:1/ 10^6 cells, $P < 0.001$). In some cases, the corresponding saturated FA-CoA was also somewhat lower (Fig. 3A). MDA-MB-231 cells also have significantly lower amounts of Cer with mono-unsaturated fatty acids (Fig. 3B) (the quantities for MDA-MB-231 versus MCF7 cells were: 24.8 ± 4.5 pmol versus 7.1 ± 1.0 pmol C24:1-Cer/ 10^6 cells, $P < 0.001$; and 3.9 ± 0.6 pmol versus 0.5 ± 0.1 pmol C26:1-Cer/ 10^6 cells, $P < 0.001$). In contrast, there were fewer differences in the Cer with saturated fatty acids (Fig. 3B). All together, these findings are consistent with the apparently lower expression of SCD in the MDA-MB-231 cells.

Likewise, the lower DES2 expression (Fig. 1) is reflected in lower amounts of phyto-sphingolipids in MDA-MB-231 cells (Fig. 4). Per 10^6 cells, the amounts in MDA-MB-231 cells (1.4 ± 0.3 pmol phyto-Cer, 0.06 ± 0.02 pmol phyto-HexCer, and 0.6 ± 0.1 pmol phyto-SM) are all lower than the amounts in MCF7 cells (6 ± 1.4 pmol phyto-Cer, $P < 0.001$; 0.21 ± 0.04 pmol phyto-HexCer, $P < 0.001$; and 17.6 ± 1.9 pmol phyto-SM, $P < 0.001$).

Shown in Fig. 5 are the data relevant to the branch point for biosynthesis of GalCer, which would be predicted to be

TABLE 2. Comparison of gene expression and glycosphingolipid differences for nine cancer datasets

Cancer Type	Gene Expression Ref.	Metabolite Ref.	G(i)M(i)	G(d)M(d)	G(i)M(d)	G(d)M(i)
MDA-MB231 / MCF7	(36)	Current analysis and (48)	4	8	4	2
IDC / normal ductal	(43)	(14, 15, 49, 50)	5		1	
Gliosarcoma	(76)	(77)	2	1	1	
GBM / white matter	(76)	(78)	1	3	2	2
Serious ovarian	(79)	(13, 80)	2			1
Mucinous ovarian	(79)	(13, 80)	4			1
SSEA3 / globo-H	(36)	(81)	3	7	5	
Melanoma cells	(16)	(16)	6	9	5	
Melanoma	(82)	(83)	3			
Total			30	28	18	6

The table compares the agreement G(i)M(i) and G(d)M(d), and disagreement G(i)M(d) and G(i)M(d), between gene expression and sphingolipid differences for nine cancer datasets. G, gene expression difference; M, metabolite difference; (i), higher ratio; and (d), lower ratio. The total for each category were used to determine the statistical significance as indicated in Fig. 8.

lower for MDA-MB-231 cells due to lower expression of UGT8, versus GlcCer, which might be higher due to the higher UGCG in MDA-MB-231 cells. Consistent with these predictions, MDA-MB-231 cells had barely detectable GalCer (0.6 ± 0.6 pmol/ 10^6 cells), whereas MCF7 cells had 9 ± 2 pmol GalCer/ 10^6 cells ($P < 0.001$). In contrast, the two cell lines contain essentially the same amounts of GlcCer (61 ± 11 pmol GlcCer/ 10^6 MDA-MB-231 cells and 66 ± 10 pmol GlcCer/ 10^6 MCF7 cells, $P = 0.41$), despite the apparent differences in UGCG expression. This might be due to consumption of GlcCer by the next step of the pathway as B4GALT6 expression also appears to be higher for MDA-MB-231 cells, and LacCer has been reported to be higher for MDA-MB-231 compared with MCF7 cells (Fig. 5, right panel) (48).

Other predictions might be made from the microarray data (Fig. 1), but they would be more tenuous. For example, there are differences in some of the CerS that imply that the chain length subspecies of MDA-MB-231 and MCF7 cells might differ (beyond what has already been addressed above); however, two of the members (CerS1 and CerS3) were not represented in this data set, which adds unknown variables. As another example, mRNAs for both SM synthases (SMS1 and SMS2) appear to be higher for MDA-MB-231 cells, but sphingomyelinases (SMPD1 and SMPD3) are also elevated, and a transporter that is thought to be involved in Cer trafficking to Golgi for SM synthesis (CERT) is lower. On the basis of the lower SM amounts in MDA-MB-231 cells (Fig. 2), it appears that one or both of the latter factors predominate(s).

Correlation of gene predictions and metabolite analysis in breast tumor tissues

Visualization of downstream steps of sphingolipid metabolism is also possible, as shown in Fig. 6 for ganglioside biosynthesis using published data from invasive ductal carcinoma (IDC) breast tumors and normal ductal tissue (43). This dataset predicts that IDC have higher expression of GM₃ synthase (which makes GM₃ from LacCer), B4GALNT1 (which adds GalNAc to LacCer to produce GA2, and to GM₃ to produce GM₂, etc., as shown), ST3gal2 (which catalyzes the transfer of sialic acid from CMP-sialic acid to GA1 to produce GM_{1b}, to GM_{1a} for GD_{1a}, etc.) and ST8SIA5 (which is responsible for biosynthesis of GT₃,

GT_{1a}, GQ_{1b}, etc.). We have not found data to evaluate all of these predictions, but published results for the sphingolipid composition of IDC tissue versus normal tissue (49) noted higher quantities of GM₃ (49) as well as higher GD₃, and N- and O-acetyl GD₃ and GT₃ (49), which might be due to elevation of GM₃ because the mRNAs for the proximal enzymes (e.g., ST8Sial and O-AcetylT in Fig. 6) did not appear to be higher than for normal tissue. It is perhaps relevant that sera from patients with IDC have been found to have significant elevations in GD_{1b}, GT_{1b}, and GQ_{1b} (highlighted in Fig. 6 as circles with colored edges) compared with healthy volunteers (15), which is also consistent with the gene expression data. It would be interesting to know whether this gene expression profile might direct investigators to additional differences in glycosphingolipid biosynthesis that might have functional relevance in this cancer or perhaps serve as biomarkers.

Fig. 7 shows a comparison of the available gene expression data for steps in the biosynthesis of globo-series glycosphingolipids for IDC versus normal tissue. From the pathway depiction, higher globoside synthase, B3GalT5, and Fut2 would be predicted to elevate Gb4, SSEA-3, and Globo-H antigen, and the latter two (highlighted in Fig. 7

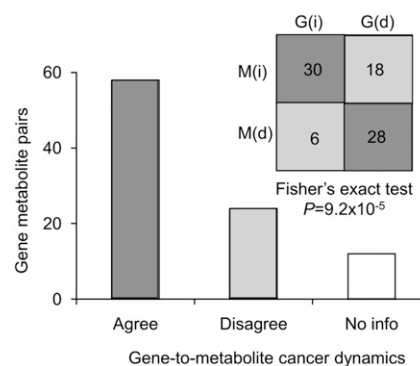


Fig. 8. Statistical comparison of gene expression differences with metabolite amounts. Results from a survey comparing nine different gene expression datasets with corresponding sphingolipids. G(i) and G(d) indicate an increase and decrease in gene levels, while M(i) and M(d) indicate increase and decrease in metabolites. The significance of agreement and disagreement between genes and metabolites was measured by Fisher's exact test ($P = 0.01$).

as circles with colored edges) have been reported to be elevated in 77% and 61%, respectively, of breast cancer tissues (14, 50).

Comparison of the frequency of gene and sphingolipid alterations in cancers

There is no comprehensive dataset that allows rigorous comparison of the correlation between gene expression data and sphingolipid composition; however, we have analyzed the frequency for the data at hand (**Table 2**): nine gene expression datasets for which corresponding metabolite data are available, for a total of 82 observations, including the present study. Evaluation of positive and negative correlations between gene expression differences and metabolite amounts revealed 30 instances in which both increased, 28 in which both decreased, for a total of 58 positive correlations (**Fig. 8** and Table 2). In 24 cases, genes and metabolite changes showed opposing outcomes (a negative correlation). In 12 instances, the metabolites were different, but the respective mRNA changes were less than ± 1.5 -fold. Thus, although we do not have data for all of the genes and metabolites of this pathway, for the ones we were able to locate there was a 70.7% correlation between gene expression and sphingolipid differences, which was statistically significant at $P < 0.001$ by Fisher's exact test (Fig. 8).

Although many factors other than mRNA amount are well known to regulate metabolic pathways (e.g., rates of translation, post translational modification, protein turnover, etc.), this degree of correlation suggests that it would still be worthwhile to use a pathway depiction of gene expression data to direct attention to particular sphingolipids for subsequent analysis by a method such as mass spectrometry. To illustrate the wide range of similarities and differences among different cancer cell lines, pathway maps based on published gene expression data (36) for the 59 cancer cell lines in the NCI 60-cell line drug screen (NCI-60, <http://discover.nci.nih.gov/cellminer/>) have been provided in the supplementary online data. There is clearly a rich amount of diversity that warrants exploration.

SUMMARY

These studies illustrate how use of pathway maps to visualize gene expression data can facilitate the prediction of metabolic changes and direct attention to particular sphingolipids for analysis. Two examples of this in our studies were the observation of elevated SPTLC3 in MDA-MB-231 cells versus MCF7 cells, which directed our attention toward the possibility that shorter chain (d16:1) sphingoid bases are present and that lower DES2 implied that MDA-MB-231 cells have less phyto-sphingolipids. In a separate study, we recently used this pathway map to note that epithelial ovarian cancer cells (collected by laser capture microdissection) displayed elevated expression of mRNAs for the steps in sulfatide biosynthesis; the presence of sulfatides was confirmed by LC ESI-MS/MS and tissue-imaging mass spectrometry (51). Considering the already large,

and continually growing, body of data on differences in gene expression in cancer and other diseases, these pathway maps should be useful in accelerating the discovery of interesting new ways that sphingolipids are involved, and possibly in identifying new disease biomarkers. As lipidomic data sets become more plentiful, there might also be occasions where the display of the metabolite data in PathVisio could assist investigators in the discovery of interesting aspects of regulation of sphingolipid metabolism at transcription or other levels.

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