Bioinformatics analysis of dysregulated microRNAs in exosomes from docetaxel-resistant and parental human breast cancer cells

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Background: Resistance to docetaxel is a major obstacle to effective treatment of breast cancer. Exosomal microRNAs (miRNAs) have recently been introduced in cell-to-cell transmission of chemoresistance between heterogeneous populations of tumor cells with diverse drug sensitivity. However, a systematic evaluation of the exosomal miRNA signature remains largely unclear.

Method: miRNA expression profiles in exosomes from docetaxel-resistant (D/exo) and parental sensitive breast cancer cells (S/exo) were assessed using microarray. Bioinformatics analysis was performed to predict target genes of the dysregulated miRNAs and to uncover their potential roles in chemoresistance formation. Signaling pathways, gene ontology terms, transcription factors, protein-protein interactions, and hub genes were also constructed.

Results: The selected exosomal miRNAs could modulate target genes responsible for MAPK, TGF-beta, Wnt, mTOR, and PI3K/Akt signaling pathways. Function enrichment analysis revealed the involvement of target genes in transcription regulation, protein phosphorylation, kinase activity, and protein binding. Enriched transcription factors including SP1, SP4, and EGR1 were obtained and a protein-protein interaction network was established. The hub genes for up-expressed and down-expressed exosomal miRNAs such as *CCND1* and *PTEN* were identified.

Conclusion: This bioinformatics study provides a comprehensive view of the function of dysregulated exosomal miRNAs, and may help us to understand exosome-mediated resistance transmission and overcome docetaxel resistance in future breast cancer therapy.

Keywords: breast cancer, exosomes, microRNA, chemoresistance, docetaxel

Introduction

Breast cancer is the most common malignant tumor and the leading cause of cancer-related death among females worldwide.¹ Docetaxel-based chemotherapy forms an important part of successful cancer treatment; however, drug resistance, whether innate or acquired over time, is almost inevitable.² Chemoresistance is a multifactorial phenomenon involving many mechanisms. Although much effort has been expended in studying the molecular basis of chemoresistance, reasons for the failure of docetaxel have yet to be fully elucidated.

Exosomes are nano-sized extracellular vesicles about 50–100 nm in diameter released from many cell types. Mounting evidence suggests that exosomes play significant roles in cancer biology, including tumorigenesis, angiogenesis,

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invasion, and metastasis.^{3,4} Added to these, exosomes are regarded as powerful transmitters of drug resistance owing to their ability to shuttle functional proteins, and especially microRNAs (miRNAs), between cells.^{5,6} Our previous group explored the miRNA expression profiles of three chemoresistant breast cancer cell lines and their corresponding exosomes, and demonstrated that a number of resistance-associated miRNAs were concentrated in isolated exosomes.⁷ Our colleagues then found that exosomes from metastatic breast cancer cells could release *miR-1246*, leading to enhanced cell proliferation, invasion, and chemoresistance in non-metastatic cancer cells.⁸ Wei et al reported that exosomal *miR-221/222* from tamoxifen-resistant breast cancer cells could increase tamoxifen resistance in recipient sensitive cells.⁹ Our previous work also demonstrated that docetaxel-resistant breast cancer cells were able to spread chemoresistance to sensitive cells by releasing exosomes and transferring specific miRNAs.¹⁰ In relation to treatment, exosomes could function as nanoparticles to deliver anti-*miR-214* to reverse chemoresistance to cisplatin in cancer cells.¹¹ All the above studies have, thus far, been conducted to support the hypothesis that exosomal miRNAs were tied to drug sensitivity and might serve as a potential alternative for breast cancer therapy.

Remarkably, most studies attempting to detect miRNA signatures relevant to chemoresistance have focused only on cellular miRNAs or on individual exosomal miRNA. A lack of knowledge regarding the miRNA expression profiles would limit the understanding of exosomal miRNA's role in chemoresistance formation. Therefore, the purpose of this work was to bioinformatically analyze the miRNA signatures in exosomes from docetaxel-resistant and parental breast cancer cells.

Materials and methods

Cell culture

Human breast cancer cell line MCF-7 was purchased from the Cell Bank of the Chinese Academy of Sciences (Shanghai, People's Republic of China). The docetaxel-resistant subline (MCF-7/Doc) was successfully established from the parental sensitive cell line (MCF-7/S) by increasing the concentration of docetaxel in our laboratory, as described previously.¹² All cell lines were cultured in DMEM high glucose (HyClone, South Logan, UT, USA) containing 10% FBS, 100 U/mL penicillin, and 100 µg/mL

streptomycin, at 37°C and 5% CO₂ in a humidified atmosphere.

Exosome isolation and characterization

Exosomes were isolated from medium of MCF-7/Doc and MCF-7/S cultured in exosome-free FBS using repeated centrifugation and ultracentrifugation steps, as described previously.¹⁰ They were respectively named D/exo and S/exo for simplicity. Characterization of harvested exosomes was performed by transmission electron microscopy as previously reported.¹³ In brief, approximately 10 µL exosome samples were placed on parafilm and covered by a 300-mesh copper grid for 45 minutes. After washing three times in PBS, the copper grid was fixed in 3% glutaraldehyde for 10 minutes, washed with double-distilled water, and then contrasted in 2% uranyl acetate. Images were captured by using a JEM-1010 electron microscope (JEOL, Tokyo, Japan) at an accelerating voltage of 80 kV.

Exosomal miRNA extraction and microarray

Exosomal RNA was extracted using the Total Exosome RNA and Protein Isolation Kit (Invitrogen, Waltham, MA, USA) in accordance with the manufacturer's protocols. RNA was quantified spectrophotometrically (Thermo Scientific, Waltham, MA, USA), and its quality was assessed by an Agilent 2100 Bioanalyzer (Agilent Technologies, Santa Clara, CA, USA). MicroRNA expression profiles were generated and hybridization using the Affymetrix GeneChip miRNA 3.0 Array was carried out as previously described.¹⁴ In brief, hybridization was performed using the Affymetrix Fluidics Station 450 (Affymetrix, Santa Clara, CA, USA) and Hybridization Oven 640. Image processing was conducted using the Affymetrix Gene Array 3000 scanner. The raw data were treated using miRNA QC tool software (Affymetrix), and data output was received in Excel spreadsheets containing the normalized miRNA expression profiles. Differentially expressed exosomal miRNAs in D/exo were filtered to exclude those changes less than 2.0-fold different compared with exosomal miRNAs in S/exo.

Real-time PCR

Real-time PCR was conducted using the SYBR PrimeScript RT-PCR kit (Takara Bio, Tokyo, Japan) on a Light Cycler 480 (Roche, Sydney, Australia) according to the

manufacturer's protocols. All reactions were performed in a 20 μ L reaction volume in triplicate. The primers for *U6* are as follows: forward, 5'CGCAAGGATGACACG3'; reverse, 5'GAGCAGGCTGGAGAA3'. The relative miRNA expressions were calculated using the $\Delta\Delta C_t$ method and normalized to *U6*.

Target gene prediction

We followed the methods of Zhong et al, as previously reported.⁷ The starBase version 2.0 (<http://starbase.sysu.edu.cn/browseClipSeq.php>) generates possible miRNA–target interactions by gathering information from five existing prediction software packages (TargetScan, PicTar, RNA22, PITA, and miRanda).¹⁵ In the present study, starBase was used to predict target genes of the selected miRNAs. Only the common target genes listed by at least four independent tools were taken into account.

KEGG pathway and GO analysis

The web-based program Database for Annotation, Visualization and Integrated Discovery (DAVID, <http://david.abcc.ncifcrf.gov/>) was used to perform functional enrichment analysis including Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway analysis and Gene Ontology (GO) annotation for the target genes.^{16–18} A count number larger than 2 and Bonferroni *P*-value <0.05 were chosen as cut-off criteria. Cytoscape software (<http://cytoscape.org/>) was applied to construct possible networks.¹⁹

Screening of transcription factors

FunRich (<http://www.funrich.org>) is a standalone functional enrichment and interaction network analysis tool that identifies the enriched transcription factors for target genes.²⁰ In the present work, FunRich was used to screen transcription factors that regulate target genes of dysregulated exosomal miRNAs.

Integration of protein–protein interaction (PPI) network

Target genes of dysregulated exosomal miRNAs were uploaded to the Search Tool for the Retrieval of Interacting Genes (STRING) database (<http://www.string-db.org/>) to evaluate PPI information, and only the interactions with a combined score >0.4 were considered as significant.²¹ The plug-in Molecular Complex Detection (MCODE) was used to screen the hub modules of the PPI network in Cytoscape with the default parameters (degree

cut-off ≥ 2 , node score cut-off ≥ 0.2 , K-core ≥ 2 , and maximum depth =100). MCODE automatically calculates the node numbers, edge numbers, and MCODE scores in the network. The degree of connectivity was also analyzed to obtain the hub genes.

Statistical analysis

Statistical analysis was performed using the SPSS 20.0 package (IBM Corp., Armonk, NY, USA). All experiments were carried out in triplicate and the data presented were representative of three independent experiments. Student's *t*-test was used to assess the statistical significance of differences between groups in PCR analysis. A value of *P*<0.05 was considered significant.

Results

Exosome characterization and miRNA profiling

D/exo and S/exo had the same morphology and size, ranging from 20 to 100 nm in diameter, as observed by transmission electron microscopy (Figure 1A and B). Expressions of small RNAs in D/exo and S/exo were screened by microarray and then categorized into miRNA, CDBox, HacaBox, scaRNA, and snoRNA. A total of 66.5% from D/exo and 52.6% from S/exo were mapped onto miRNAs. Compared to S/exo, 208 and 99 miRNAs were significantly up-expressed and down-expressed (at least 2.0-fold changes), respectively, in D/exo. In all the differentially expressed miRNAs, D/exo had 44 novel miRNAs with increased expression levels >16.0-fold and five novel miRNAs with reduced expression levels >16.0-fold of the corresponding miRNAs in S/exo (Table 1). Hierarchical clustering analysis of miRNA signal intensities showed evidence of significant changes of miRNA expression in D/exo and S/exo (Figure 1C). The miRNA microarray data were further validated by PCR using the following selected exosomal miRNAs (*miR-23a*, *miR-27a*, *miR-30a*, and *miR-222*), and all trends of miRNA expression by PCR coincided with microarray (not shown).

Target gene prediction and pathway analysis

From the top 10 most differentially expressed miRNAs in D/exo, the nine most up-expressed miRNAs (*let-7a*, *let-7b*, *let-7c*, *miR-103a*, *miR-16*, *miR-23a*, *miR-23b*, *miR-27a*,

Figure 1 Exosome characterization and miRNA profiling. Representative transmission electron microscopy images of D/exo (A) and S/exo (B), showing the same morphology and a diameter of 20–100 nm (scale bar, 200 nm). (C) Hierarchical cluster analysis of miRNA expression profiles in D/exo with respect to S/exo. Yellow and blue colors stand for a transcript level above and below the median level, respectively.

Abbreviations: D/exo, exosomes from docetaxel-resistant cells; S/exo, exosomes from parental sensitive cells.

and *miR-30a*) and eight most down-expressed miRNAs (*miR-130a*, *miR-20b*, *miR-25*, *miR-425*, *miR-455-3p*, *miR-4725-5p*, *miR-551*, and *miR-92*), which could be found in the starBase database, were selected for evaluation. The starBase tool was used for prediction, generating 1,857 potential target genes, including 953 genes for up-expressed miRNAs and 904 genes for down-expressed miRNAs.

To explore which pathways might be involved in chemoresistance formation, the predicted genes were assigned into the KEGG pathway, and Cytoscape software was applied to decipher the possible network. The enriched pathways for target genes of up-expressed miRNAs included “MicroRNAs in cancer”, “Signaling pathways regulating pluripotency of stem cells”, “MAPK signaling pathway”, “TGF-beta signaling pathway”, “FoxO signaling pathway”, and “Wnt signaling pathway”. For down-expressed miRNAs, the enriched pathways included “MAPK signaling pathway”, “TGF-beta signaling pathway”, “Pathways in cancer”, “FoxO signaling pathway”, “mTOR signaling pathway”, “PI3K/Akt signaling pathway”, and “Proteoglycans in cancer” (Figure 2). In breast

cancer, these signaling pathways have been found to participate in drug resistance and treatment failure.^{22–26}

GO enrichment and transcription factor analysis

To explore the role that predicted target genes play in biological functions, GO enrichment analysis was performed. The results showed that the selected miRNAs were significantly enriched in biological processes, including “protein phosphorylation/dephosphorylation”, “positive/negative regulation of transcription, DNA-templated”, “positive/negative regulation of cell proliferation”, and “positive/negative regulation of transcription from RNA polymerase II promoter” (Figure 3). For molecular function, the selected miRNAs were mainly related to “protein binding”, “kinase binding”, “DNA binding”, “ATP binding”, and “kinase activity” (Figure 4). In addition, cellular component analysis indicated that the selected miRNAs were involved in “cytoplasm”, “nucleoplasm”, “nucleus”, “membrane”, “cytosol”, and “cell–cell adherens junction” (Figure 5).

Table 1 Novel miRNAs with expression levels >16.0-fold in D/exo with respect to S/exo

Up-expressed exosomal miRNAs	Log2 (fold change) *	Down- expressed exo- somal miRNAs	Log2 (fold change)
<i>hsa-let-7a_st</i>	8.50	<i>hsa-miR-1263_st</i>	-4.13
<i>hsa-let-7b_st</i>	9.01	<i>hsa-miR-1281_st</i>	-4.14
<i>hsa-let-7c_st</i>	8.04	<i>hsa-miR-3162-3p_st</i>	-4.15
<i>hsa-let-7d_st</i>	4.06	<i>hsa-miR-3613-5p_st</i>	-5.93
<i>hsa-let-7e_st</i>	6.38	<i>hsa-miR-455-3p_st</i>	-5.69
<i>hsa-let-7f_st</i>	6.01		
<i>hsa-let-7i_st</i>	5.55		
<i>hsa-miR-103a_st</i>	6.83		
<i>hsa-miR-107_st</i>	6.49		
<i>hsa-miR-1246_st</i>	9.60		
<i>hsa-miR-125a- 5p_st</i>	4.06		
<i>hsa-miR-125b_st</i>	5.57		
<i>hsa-miR-126_st</i>	4.42		
<i>hsa-miR-1290_st</i>	5.72		
<i>hsa-miR-155_st</i>	6.71		
<i>hsa-miR-15b_st</i>	5.96		
<i>hsa-miR-16_st</i>	8.13		
<i>hsa-miR-17_st</i>	4.60		
<i>hsa-miR-181a_st</i>	5.64		
<i>hsa-miR-181b_st</i>	5.22		
<i>hsa-miR-182_st</i>	4.31		
<i>hsa-miR-185_st</i>	4.36		
<i>hsa-miR-18a_st</i>	4.52		
<i>hsa-miR-191_st</i>	5.16		
<i>hsa-miR-193b_st</i>	5.48		
<i>hsa-miR-19b_st</i>	5.57		
<i>hsa-miR-20b_st</i>	4.55		
<i>hsa-miR-21_st</i>	4.36		
<i>hsa-miR-221_st</i>	5.24		
<i>hsa-miR-222_st</i>	5.49		
<i>hsa-miR-23a_st</i>	9.88		
<i>hsa-miR-23b_st</i>	7.72		
<i>hsa-miR-25_st</i>	4.29		
<i>hsa-miR-26a_st</i>	6.58		
<i>hsa-miR-27a_st</i>	7.39		
<i>hsa-miR-27b_st</i>	6.26		
<i>hsa-miR-29a_st</i>	6.23		
<i>hsa-miR-30a_st</i>	7.21		
<i>hsa-miR-30c_st</i>	4.62		
<i>hsa-miR-3135b_st</i>	4.69		
<i>hsa-miR-3188_st</i>	4.20		
<i>hsa-miR-31_st</i>	5.70		
<i>hsa-miR-92a_st</i>	5.25		
<i>hsa-miR-93_st</i>	4.15		

Note: *Fold change = (D/exo)/(S/exo).

Abbreviations: D/exo, exosomes from docetaxel-resistant cells; S/exo, exosomes from parental sensitive cells.

Based on data obtained from FunRich, the top 10 enriched transcription factors for the target genes of up-expressed miRNAs were SP1, SP4, KLF7, EGR1, HNF4A, POU2F1, ETS1, NF1C, NFYA, and MEF2A (Figure 6A). For the target genes of down-expressed miRNAs, the top 10 enriched transcription factors were SP1, SP4, EGR1, KLF7, POU2F1, ETS1, NFYA, NFIC, RREB1, and MEF2A (Figure 6B).

PPI construction and selection of hub genes

According to the information obtained from STRING database, target genes of the selected miRNAs interacted with each other, and the PPI network consisted of 1,414 nodes and 12,272 edges. For better visualization, the top three modules with the top three highest MCODE scores were selected using the plug-in MCODE (Figure 7A–C). Enrichment pathway analysis indicated that the genes in the modules were mainly associated with “PI3K/Akt signaling pathway” and “mTOR signaling pathway” (Figure 7B and C). For the up-expressed miRNAs, the hub genes were *CCND1*, *PTEN*, *BCL2*, *KRAS*, *MAP2K1*, *BTRC*, *XPO1*, *ACTA1*, *NRAS*, and *CREB1*. *CCND1* showed the highest node degree (degree = 86). For the down-expressed miRNAs, the hub genes were *PTEN*, *CCND1*, *CALM2*, *CDKN1A*, *STAT3*, *RB1*, *ITCH*, *ACTR1A*, *SP1*, and *CNOT6L*. *PTEN* exhibited the highest node degree (degree = 79) (Table 2).

Discussion

Resistance to docetaxel is a major barrier for successful treatment of breast cancer. Understanding the molecular machinery is of critical importance for reversing drug insensitivity.² Exosomal miRNAs have been confirmed to play important roles in the development of chemoresistance.^{4,5} Our previous group explored the miRNA expression profiles of three breast cancer cell lines, respectively resistant to docetaxel, epirubicin, and vinorelbine, and their exosomes, and demonstrated that a number of resistance-associated miRNAs were concentrated in the corresponding exosomes.⁷ Moreover, we confirmed that docetaxel-resistant breast cancer cells could transmit resistance capacity through altering gene expression in sensitive cells by transferring specific miRNAs contained within exosomes.¹⁰ It was also reported by our team that exosomal *miR-1246* could promote cell proliferation, invasion pathways, and drug resistance in breast

Figure 2 Analysis of involved pathways. KEGG pathway analysis of target genes of the nine most up-expressed miRNAs and eight most down-expressed miRNAs in D/exo cells showed a functional network.

Abbreviations: KEGG, Kyoto Encyclopedia of Genes and Genomes; D/exo, exosomes from docetaxel-resistant cells.

cancer.⁸ Therefore, the differential exosomal miRNA expression could partly explain docetaxel resistance in breast cancer cells. In the current study, bioinformatics analysis of miRNA signatures in exosomes from docetaxel-resistant and parental MCF-7 breast cancer cells was performed to investigate key pathways and biological functions associated with chemoresistance. To the best of our knowledge, this is the first report of integrative bioinformatics analysis of the dysregulated exosomal miRNAs in docetaxel-resistant MCF-7 breast cancer cells.

The microarray is increasingly valued as a promising tool in medical oncology, with many applications. In our previous study, we screened differentially expressed miRNAs between docetaxel-resistant and sensitive breast cancer cells.¹⁴ In the present work, miRNA expression profiles of D/exo and S/exo were also analyzed by microarray to explore which miRNAs might be responsible for

exosome-mediated resistance transmission. With respect to S/exo, a total of 208 and 99 miRNAs were up-expressed and down-expressed in D/exo, indicating that D/exo were characterized by significant changes in miRNA expression and were miniature maps of their cells of origin. This result, along with our previous observation, is consistent with that of Zhong et al, who explored the miRNA expression profiles of three drug-resistant breast cancer cell lines and their exosomes and demonstrated that a number of resistance-associated miRNAs were concentrated in the isolated exosomes.⁷

During bioinformatics analysis, we followed the methods as previously described.⁷ KEGG pathway analysis was performed to check potentially involved pathways of target genes of the nine most up-expressed miRNAs and eight most down-expressed miRNAs in D/exo. From the results, a single exosomal miRNA could influence multiple pathways

Figure 3 Analysis of biological processes. Gene Ontology (GO) enrichment analysis of biological processes of the target genes showed a functional network.

Figure 4 Analysis of molecular function. Gene Ontology (GO) enrichment analysis of molecular functions of the target genes showed a functional network.

Figure 5 Analysis of cellular components. Gene Ontology (GO) enrichment analysis of cellular components of the target genes showed a functional network.

Figure 6 Analysis of transcription factors. The top 10 most enriched transcription factors for the target genes of (A) up-expressed and (B) down-expressed exosomal miRNAs are presented

(eg, *miRNA-16* and *miRNA-20b*) and the same pathway could be regulated by multiple miRNAs. Several classical signaling pathways associated with chemoresistance were identified from the top enriched KEGG terms, namely MAPK, TGF-beta, Wnt, mTOR, and PI3K/Akt signaling pathways. Breast cancer arises from a subpopulation of malignant cells with high tumor initiation and self-renewal capability, known as cancer stem cells, which are implicated to be a cause of metastasis, tumor relapse, and therapy resistance.²⁷ The

enriched pathways for target genes of both up-expressed exosomal miRNAs and down-expressed miRNAs contained “FoxO signaling pathway”. FoxO gene is considered to limit cell proliferation and induce apoptosis, serving as a mediator of tumor response to various therapies.²⁴ Therefore, understanding the target genes involved in “signaling pathways regulating the pluripotency of stem cells” and “FoxO signaling pathway” may help to uncover the mechanisms of docetaxel resistance.

Figure 7 Analysis of the protein-protein interaction network. **(A)** Module 1 and the enriched pathways. **(B)** Module 2 and the enriched pathways. **(C)** Module 3 and the enriched pathways.

GO annotation and enrichment analysis demonstrated that the target genes of selected exosomal miRNAs were significantly enriched in functions involving transcription regulation, protein phosphorylation, kinase activity, and protein binding. Numerous studies have confirmed that defects of biological processes, molecular functions, and cellular components would influence cancer biology, promote cell proliferation, invasion, and metastasis, and change chemosensitivity.²⁸ How these GO terms are implicated in drug resistance needs further investigation.

SP1, SP4, and EGR1 were the common enriched transcription factors for the target genes of up-expressed and down-expressed exosomal miRNAs. There is evidence

that transformation of normal cells to malignant cells is associated with increased SP1 and SP4, which are highly expressed in cancer and responsible for cell growth, survival, migration, and drug resistance.²⁹ A 2016 study indicated that EGR1 is an important regulator in a triple-negative breast cancer cell line and is a promising marker for docetaxel resistance.³⁰ Therefore, exploring the possible transcription factors may represent another useful way toward the understanding of chemoresistance.

A PPI network was constructed and it was found that *CCND1* and *PTEN* were the hub genes with the highest node degree among the targets of up-expressed and down-expressed exosomal miRNAs, respectively. *CCND1* is

Table 2 Hub genes identified in the protein–protein interaction network

Up-expressed exosomal miRNAs		Down-expressed exosomal miRNAs	
Gene symbol	Degree	Gene symbol	Degree
<i>CCND1</i>	86	<i>PTEN</i>	79
<i>PTEN</i>	79	<i>CCND1</i>	75
<i>BCL2</i>	76	<i>CALM2</i>	67
<i>KRAS</i>	75	<i>CDKN1A</i>	63
<i>MAP2K1</i>	67	<i>STAT3</i>	55
<i>BTRC</i>	66	<i>RBI</i>	52
<i>XPO1</i>	64	<i>ITCH</i>	50
<i>ACTA1</i>	64	<i>ACTR1A</i>	48
<i>NRAS</i>	62	<i>SPI</i>	46

required for the transition of cells from G1 to S phase and is overexpressed in up to 50% of human breast cancers.³¹ *CCND1* orchestrated expression of a miRNA signature that induced the Wnt signaling pathway, serving both upstream and downstream of the Wnt cascade.³² Given the emerging nature of this field, we are currently trying to determine the association between exosomal miRNAs, *CCND1*, and the Wnt signaling pathway, and to verify their effects using gain- and loss-of-function assays. *PTEN* has emerged as the key functional antagonist to PI3K, controlling activation of the PI3K/AKT/mTOR signaling pathway. According to a 2017 meta-analysis, *PTEN* loss was of particular significance for predicting breast cancer aggressiveness and poor outcome.³³ Our previous studies demonstrated that *miR-222* could decrease the expression of *PTEN*, through which *miR-222* may confer chemoresistance in breast cancer cells.¹⁴ In addition, we have confirmed that D/exo from docetaxel-resistant breast cancer cells were able to down-regulate *PTEN* expression.¹⁰ Analysis of the top three highest modules revealed the involvement of the genes in the PI3K/Akt signaling pathway and the mTOR signaling pathway. These results support that *PTEN* and the PI3K/Akt signaling pathway may be central points in exploring exosomal miRNA's role in chemoresistance formation. More experiments are necessary to support this conclusion.

Conclusions

In summary, our work provides a comprehensive bioinformatic analysis of the dysregulated miRNAs in exosomes from docetaxel-resistant and parental sensitive breast cancer cells. After target gene prediction, KEGG pathways,

GO terms, transcription factors, PPI network, and hub genes were evaluated. Further laboratory studies are needed to more precisely investigate the functions of target genes and pathways of exosomal miRNAs in the formation of chemoresistance.

Data availability

The data sets analyzed during the current study are available from the corresponding author on reasonable request.

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Author contributions

All authors contributed to data analysis, drafting and revising the article, gave final approval of the version to be published, and agree to be accountable for all aspects of the work.

Disclosure

The authors report no conflicts of interest in this work.

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