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LDLR-mediated lipidome–transcriptome reprogramming in cisplatin insensitivity

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Abstract

Platinum-based therapy remains the cornerstone for cancer therapy; however, its efficacy varies. The role of lipoprotein receptor-mediated lipid entry for cancer development has been reported. Yet, the roles and mechanism of the low-density lipoprotein receptor (LDLR) in chemo-sensitivities are unknown. In the current report, we used epithelial ovarian cancer (EOC), composed of various cellularities, to study this issue. Using public cDNA microarray database and single cohort study, LDLR expressions were positively associated with epithelial ovarian carcinomas (EOCs) platinum-based chemotherapy patients' disease prognosis. In vitro and in vivo add-in/silencing LDLR was introduced to determine cisplatin sensitivity and cancer growth. Results revealed that knocked-down LDLR could sensitize while overexpressed LDLR could insensitize EOC cells to the cytotoxic effects of cisplatin. Moreover, the trans-omics approaches depicted an LDLR→LPC (Lyso-phosphatidylcholine)→FAM83B (phospholipase-related)→FGFRs (cisplatin sensitivity and phospholipase-related) regulatory axis. Finally, the manipulation of LDLR expression in EOC cells was found to determine the efficacy of cisplatin therapy in terms of tumor suppression. In conclusion, the LDLR→LPC→FAM83B→FGFRs axis is an example of tumor macroenvironmental regulation of therapy outcomes. Relatedly, LDLR expression could serve as a biomarker of chemotherapy sensitivity in EOCs.

Key Words
- LDLR
- lysophosphatidylcholine
- EOC
- trans-omics
- bioinformatics

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Introduction

Platinum-based adjuvant chemotherapy is widely used in the management of patients with solid tumors, including gynecological malignancies (Lawrie et al. 2015, Falcetta et al. 2016), gut cancers (Malka et al. 2014, Galdy et al. 2016, Ishikawa et al. 2016), urinary tract carcinomas (Roupret et al. 2016), non-small-cell lung cancers (de Castria et al. 2013, Berghmans et al. 2017), and many other malignancies. Platinum can form DNA adducts in fast-growing cancer cells, a capacity which makes it an excellent tumor growth-suppressing agent. However, while platinum-based chemotherapies are commonly used to treat human malignancies, variations in the responsiveness and resistance of patients to such therapies are also seen (Johnson et al. 2011, Bellmunt et al. 2013, Poonawalla et al. 2015). The mechanisms underlying platinum-based chemotherapy sensitivity and resistance are yet to be elucidated; however, there is an interesting correlation between histological cellularity and platinum therapy responsiveness (Glasspool & McNeish 2013, Falcetta et al. 2016, Ferreira et al. 2016, Prendergast et al. 2017).

Among various solid tumors, epithelial ovarian carcinomas (EOCs) appear to arise from diverse origins of epithelium including ovarian epithelial cells, the fallopian tubes, and cells that have migrated from the endometrium or the intestines (Vargas 2014, Klotz & Wimberger 2017). Despite inconclusive evidence regarding the origins of EOCs, however, investigating the chemotherapy sensitivity of EOC cells might clarify why variations in platinum responsiveness occur among EOCs (McCluggage 2011). The different subtypes of EOCs can be classified histologically as serous, mucinous, endometrioid, and clear-cell EOCs (Vargas 2014), and previous studies have found that these subtypes of EOCs are associated with varying levels of platinum-based therapy efficacy (Glasspool & McNeish 2013, Ledermann et al. 2014, Lawrie et al. 2015, Matulonis et al. 2016, Dahm-Kahler et al. 2017). Serous EOC cells/patients have good sensitivity to chemotherapy (McCluggage 2011). In contrast, advanced clear-cell EOCs have been found to exhibit poor responses to platinum-based chemotherapy, with recurrent clear-cell carcinomas appearing to be particularly resistant to chemotherapy and difficult to treat (Sugiyama et al. 2000, Mizuno et al. 2006, Lee et al. 2011). Malignant mucinous tumors are epithelial ovarian tumors formed by cells that resemble the endocervical epithelium or intestinal epithelium (intestinal type), and the efficacy of platinum-based therapies against these tumors has also been found to be inferior (Ricci et al. 2018). Meanwhile, the clinicopathological features of malignant endometrioid EOCs are similar to those of clear-cell EOCs, as these subtypes are postulated to arise from the same cell type (Shevchuk et al. 1981) and to be platinum-based chemotherapy insensitive (Sugiyama et al. 2000).

Lipids are essential for biomass and building block synthesis, and lipids can also act as bioactive molecules, for example, as constituents of cellular membranes, or as an energy supply sufficient for the fast-growing nature of cancer cells (Ward & Thompson 2012). Microscopically, endometrioid EOC exhibits an appearance similar to that of tubular glands and bears a resemblance to endometrium. Squamous differentiation is commonly seen in endometrioid EOC patients (Wagner et al. 1994), along with surrounding lipid droplet (LD)-vacuolated stromal tissue (Ulbright & Roth 1985). In addition, clear-cell EOC is also characterized by a significant amount of LD vacuoles in the cytoplasm (Nishimura et al. 2010). Moreover, the lipophilic nature of metastatic EOC cells appears to make them keen to migrate to the omentum (Nieman et al. 2011). All of those characteristics have raised the question of whether non-autonomous lipid providers, i.e. tumor macroenvironmental regulators (Lee et al. 2016) produced via the lipoprotein-mediated lipid route (lipoprotein/receptor-route) (Chang et al. 2017), might play roles in the development and disease progression of EOCs.

In the current study, we discovered that different levels of low-density lipoprotein receptor (LDLR; the gateway for non-autonomous lipid entry) expression in EOCs determine the platinum sensitivity of EOCs in an LDLR-dependent manner. In addition, LDLR alters the lipid and gene expressions of EOC platinum therapy sensitivity.

Materials and methods

Patient cohort

The paraformaldehyde-embedded EOC tissue samples analyzed in this study were obtained from patients diagnosed with EOC from 2008 to 2013 at China Medical University (Taichung, Taiwan). The patients were identified from a single cohort registered in the Cancer Registry Database of the hospital, and the gynecological pathology of each patient was classified according to the World Health Organization pathology classification. Access to the tissue samples was approved by the Internal
Review Board of China Medical University Hospital (#DMR101-IRB2-276 and CMU105-REC3-122(CR1)). Consent was obtained from each patient or subject after a full explanation of the purpose and nature of all the procedures used. The EOC subtypes were selected through a patient chart review and then confirmed based on the analysis of hematoxylin and eosin (H&E)-stained and sectioned paraffin slides by two pathologists to exclude ambiguity or mixed histology between subtypes.

Immunohistochemistry and quantitation of staining score

In general, the histological studies were performed as described in previous studies (Hung et al. 2014, Chen et al. 2018) with some modifications. For general histologic inspection, we treated the tissue sections (2 μM) with H&E or stained the sections with antibodies specific for LDLR and lipoprotein lipase (LPL) with an ABC kit (Vector Laboratories) to enhance the staining signals. Staining intensity was scored according to the Allred scoring system (Nose et al. 2009, Hammond et al. 2010) and our previous work (Lai et al. 2016). The proportion of cells that stained positive for LDLR and LPL was graded using a five-point scale according to the proportion of positive cancer cells (1: <1/100; 2: 1/100 to 1/10; 3: 1/10 to 1/3; 4: 1/3 to 2/3; and 5: >2/3). The intensity of staining was also graded on a five-point scale as follows: 1: none; 2: weak; 3: intermediate; 4: mid-strong; 5: strong. The proportions and intensity scores were next summed, averaged, and then compared with the associated histological reports. The slides were independently examined by three coauthors (WC Chang, MD; HS Wang; and YP Ho) who were blinded to the clinicopathological data. When there was a discrepancy (i.e., a score difference >3) between the scores given by the slide reviewers, the pathologists reassessed the slides using a double-headed microscope, and a consensus was reached. Finally, associations between the scores and the clinical data were investigated by another coauthor (YC Hung, MD).

Reagents, cell culture, and lentiviral-based gene delivery

Cells were maintained in various culture media (depending on the culture requirements) with 10% FCS (Invitrogen), 1% L-glutamine, and 1% penicillin/streptomycin as described previously (Chang et al. 2017). The HEK293T (ATCC; HTB52) and EOC (MDAH-2774; SKOV3, HTB-77; OVCA3, HTB-161; ES2, CRL-1978; TOV-112D, CRL-11731; TPOV-21G, CRL-11730) cell lines were purchased from ATCC.

The following antibodies were used: LDLR (for IHC: GeneTex, GTX61553; for immunoblot: Santa Cruz, sc-373830), LPL (Santa Cruz, sc-32885), VEGFR (Santa Cruz, sc-6251), phospho-FAK (Cell Signaling Technology, #3283; Try397), FAK (Cell Signaling Technology, #3285), MEK1/2 (Cell Signaling Technology, #8727), actin (Santa Cruz, sc-47778), and tubulin (Abcam, ab-6046). The following chemicals were also used: cisplatin (Sigma-Aldrich, P4394), DOTAP (1,2-dioleoyl-3-trimethylammonium-propane; Avanti, 890890P), and lysophosphatidylcholine (Avanti, 855675P).

Lentiviral-based gene delivery

LDLR knockdown or overexpression clones were engineered by the stable transfection of human LDLR cDNA (pLenti-C-mGFP-LDLR, RC200006L2; OriGene, Rockville, MD, USA) or pLKO.1-shLDLR (targeting sequence shown in Supplementary Table 1, see section on supplementary materials given at the end of this article) and then selected after exposure to puromycin (10 μM) for a period of time (Ma et al. 2012, 2014). The pLKO-shLuciferase, shLDLR, shFAM83B (Supplementary Table 2) plasmids were obtained from the National RNAi Core Facility Platform (Institute of Molecular Biology/Genome Research Center, Academia Sinica, supported by the National Core Facility Program for Biotechnology; grant NSC107-2319-B-001-002). The pBabe and pWPI (Addgene) vector-based LDLR cDNA expression plasmids were constructed as previously reported (Ma et al. 2008). The lentiviral production and infection procedures used in this study followed those reported previously (Ma et al. 2012). In brief, psPAX2 (packaging plasmid) and pMD2G (envelope plasmid) (Addgene) were co-transfected into HEK293T cells. We then harvested virus-containing media to infect the EOC cells. The GFP+ cell populations, as determined by flow cytometry analysis (BD LSR II Flow Cytometry), were used to test the infection efficiencies.

Colonies formation, cytotoxic measurement, and IC 50 values

Colonies were maintained in various culture media (depending on the culture requirements) with 10% FCS (Invitrogen), 1% L-glutamine, and 1% penicillin/streptomycin as described previously (Chang et al. 2017). The HEK293T (ATCC; HTB52) and EOC (MDAH-2774; SKOV3, HTB-77; OVCA3, HTB-161; ES2, CRL-1978; TOV-112D, CRL-11731; TPOV-21G, CRL-11730) cell lines were purchased from ATCC.

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with crystal violet for 5 min. After being washed with PBA, the colonies were photographed.

For the cell viability assay, cells were seeded in 96-well plates (5 x 10³ cells/well) and incubated overnight for attachment, and were then treated with the indicated doses of drugs in normal media for 48 h. After the treatments, the media were replaced with WST-1 (0.5 mg/mL) at 37°C for at least 1 h. After the removal of excess WST-1 (Sigma-Aldrich), the colorimetric absorbance of the cells at 490 nm was read. The readings of the measured values of 50% inhibition concentration (IC50) (Chou 2010) for each drug were determined by CalcuSyn software (Chou & Talalay 1984) (BioSoft).

**Experimental animal and xenograft implantation tumor model**

Athymic nude female mice aged 6–8 weeks old (Foxn1nu) were purchased from the National Laboratory Animal Center (NLAC), Taiwan. Subcutaneous implantation of 1 x 10⁶ cells/100 µL PBS and matrigel (1:1) in both flanks was performed on each mouse. The mice were then randomly divided into experimental groups as the tumors grew to 500-mm³, and the size of each tumor was measured twice/week. The mice were treated with/without cisplatin (6 mg/kg/mice or equal volume of phosphate buffered saline; I.P., every other day for 4 weeks). The mice were then killed and the tumors were harvested. All the animal studies were performed under the supervision, guidelines, and approval of the China Medical University Animal Care and Use Committee (#CMOICUC-2018-089).

**Real-time RT-PCR and primers**

The protocol for detecting mRNA expression followed that detailed in a previous publication (Chiang et al. 2017) with some modifications. Total RNA was isolated from the tissue using the Trizol™ reagent (Invitrogen) according to the manufacturer’s protocol. The mRNA levels of various genes were measured by qPCR using the Bio-Rad CFX 96 sequence detection instrument. The levels of mRNA were normalized with GAPDH mRNA. The SYBR probe (Bio-Rad) was used as the fluorogenic probe to determine the threshold cycle (Ct), and the forward and reverse primers are shown in Supplementary Table 1.

**Lipid profiling for lipidome analysis**

After the cells (1500 cells/µL x 300 µL) were washed with Ca²⁺/Mg²⁺-free PBS, the lysates were then subjected to lipid profiling executed by Lipotype GmbH (Dresden, Germany) (Ejsing et al. 2009, Sampaio et al. 2011, Levental et al. 2017). Lipidomes were prepared from at least three replicates of each sample for all the experiments using the following procedures.

**Nomenclature**

The following lipid names and abbreviations were used: Cer, ceramide; Chol, cholesterol; DAG, diacylglycerol; HexCer, glucosyl/galactosyl ceramide; PA, phosphatidic acid; PC, phosphatidylcholine; PE, phosphatidylethanolamine; PG, phosphatidylglycerol; PI, phosphatidylinositol; PS, phosphatidylserine; and their respective lysospecies lyso-PA (LPA), lyso-PC (LPC), lyso-PE (LPE), lyso-PI (LPI), and lyso-PS (LPS); and their ether derivatives PC-O-, PE-O-, LPC-O-, and LPE-O-; SE, sterol ester; SM, sphingomyelin; SLs, sphingolipids; and TAG, triacylglycerol. Lipid species were annotated according to their molecular composition as follows: (lipid class)-(sum of carbon atoms in the fatty acids):(sum of double bonds in the fatty acids):(sum of hydroxyl groups in the long-chain base and the fatty acid moiety) (for example, SM-32:2;1). Where available, individual fatty acid compositions following the same rules were given in brackets (for example, 18:1;0-24:2;0).

**Lipid extraction for MS lipidomics**

Lipids were extracted using a two-step chloroform/methanol procedure. Samples were spiked with an internal lipid standard mixture containing cardiolipin (CL), 16:1/15:0/15:0/15:0; Cer, 18:1/2;17:0; DAG, 17:0/17:0; HexCer, 18:1/2;12:0; LPA, 17:0; LPC, 12:0; LPE, 17:1; LPG, 17:1; LPI, 17:1; LPS, 17:1; PA, 17:0/17:0; PC, 17:0/17:0; PE, 17:0/17:0; PG, 17:0/17:0; PI, 16:0/16:0; PS, 17:0/17:0; cholesterol ester (CE), 20:0; SM, 18:1/2;12:0;0; TAG, 17:0/17:0; and Chol. After extraction, the organic phase was transferred to an infusion plate and dried in a speed vacuum concentrator. Each first-step dry extract was resuspended in 7.5 mM ammonium acetate in chloroform/methanol/propanol (1:2:4, v/v/v), and each second-step dry extract was resuspended in a 33% ethanol solution of methylamine/chloroform/methanol (0.003:5:1, v/v/v). All liquid handling steps were performed using the Hamilton Robotics STARlet robotic platform with the Anti-Droplet Control feature for organic solvent pipetting.

**MS data acquisition**

Samples were analyzed by direct infusion on a Q-Exactive mass spectrometer (Thermo Scientific) equipped with
a TriVersa NanoMate ion source (Advion Biosciences). Samples were analyzed in both the positive and negative ion modes with a resolution of 280,000 at m/z = 200 for MS and 17,500 for MS/MS experiments in a single acquisition. MS/MS was triggered by an inclusion list encompassing corresponding MS mass ranges scanned in 1-Da increments. Both MS and MS/MS data were combined to monitor CE, DAG, and TAG ions as ammonium adducts; PC and PC O- as acetate adducts; and CL, PA, PE, PE O-, PG, PI, and PS as deprotonated anions. Only MS was used to monitor LPA, LPE, LPE O-, LPI, and LPS as deprotonated anions; Cer, HexCer, SM, LPC, and LPC O- as acetate adducts; and Chol as an ammonium adduct of an acetylated derivative (Surma et al. 2015).

**Data analysis and post-processing**

Lipid identification using LipotypeXplorer (2) was performed on unprocessed mass spectra. For the MS-only mode, lipid identification was based on the molecular masses of the intact molecules. The MS/MS mode included the collision-induced fragmentation of lipid molecules, and lipid identification was based on both the intact masses and the masses of the fragments. Prior to normalization and further statistical analysis, the lipid identifications were filtered according to mass accuracy, occupation threshold, noise, and background (Klose et al. 2013). Lists of identified lipids and their intensities were stored in a database optimized for the particular structure inherent to lipidomic datasets. The intensity of lipid class-specific internal standards was used for lipid quantification (Liebisch et al. 2006). The identified lipid molecules were quantified by normalization to a lipid class-specific internal standard. The amounts in p-moles of individual lipid molecules (species of subspecies) of a given lipid class were summed to yield the total amount of the lipid class. The amounts of the lipid classes may be normalized to the total lipid amount, yielding mol. % per total lipids.

**Lipidomic data processing**

The lipid profiling data for each sample were scale normalized by the total amount of lipid. Those lipids with at least a two-fold change between the LDLR knockdown and control cells were identified as the lipids that were significantly regulated by LDLR. Then, Fisher’s exact test was conducted to test the enrichment of the significantly regulated lipids in each lipid class (such as PC, PE, and LPC).

**TCGA-database DriverDB.v2 and KMplotter meta-analysis for cancer survival analysis, and bioinformatics for trans-omics approach**

**Cancer survival score**

The previously developed DriverDB (Cheng et al. 2014, Chung et al. 2015), a database that incorporates more than 9500 cancer-related RNA-seq datasets and more than 7000 exome-seq. datasets from The Cancer Genome Atlas (TCGA), was used in this study. In the DriverDB, there are 420 primary tumors and 37 adjacent normal tissues (including 34 normal-tumor pairs) in the EOC dataset from TCGA. For the survival analysis of TCGA data, Kaplan–Meier survival curves were drawn and the log-rank test was performed to assess the differences between the patient groups stratified by the median of gene expression. A P value of less than 0.05 was considered statistically significant.

With regard to the Web-based KMplotter platform used for the evaluation of the hazard ratio (HR) scores of the pathways (cluster of genes) with respect to patient survival, the following previously established formula was used (Chang et al. 2016, 2017):

\[
HR = \frac{\text{Avg. of HR of gene sets}}{\text{Σ(HR}−1\text{)}×(-\log_{10}(P\text{-value})) × 100}
\]

In order to evaluate the impact of each gene, the absolute value of the HR for that gene minus 1 was calculated. To adjust for the effects of the genes, the HR value for each gene was multiplied by negative \(\log_{10}(P\text{-value})\) to balance the importance of the genes. The summed score was then divided by the number of genes and multiplied by 100 to obtain the HR score, or the average HR of all the genes.

**Trans-omics approach**

The raw reads of RNA-seq data were aligned with TopHat 2.0.13 (Kim et al. 2013) to human genome (GRCh38) with default parameters and then were assembled by Cufflink 2.2.1 (Trapnell et al. 2012), using Ensembl v90 annotations. The abundance of gene expression was calculated by fragments per kb of exon per million fragments mapped (FPKM). The normalized expression of lipidome data was obtained from Lipotype Gmbh. The differentially expressed (DE) genes/lipid species between the knockdown and control cell lines were identified by the criterion: \(|\log2(\text{Fold change})| > 1\). For the DE genes, we performed functional enrichment analysis, as described in our previous studies (Cheng et al. 2014,
Chung et al. 2016), to interpret their biological functions. In brief, we used the topGO and GeneAnswers packages of Bioconductor to calculate the topology of the GO graph. We used collections from KEGG (Kanehisa et al. 2012), PID (Schafer et al. 2009), Biocarta (http://www.biocarta.com/), REACTOME (Croft et al. 2011), and MSigDB (Subramanian et al. 2005) to annotate the DE genes in pathway level. For the DE lipid species, we performed Fisher’s exact test to identify the significant lipid classes with enriched DE lipid species. Finally, we focus on the DE genes related to the significant lipid class according to the information GO, KEGG and REACTOME database.

Preparation and characterization of LPC liposome

The liposome was prepared by thin-layer hydration (Jang et al. 2013) followed by application of the membrane protrusion method (Ong et al. 2016) with some modifications. First, we hydrated a mixture of DOTAP (mw = 698.5 g) or LPC (mw = 495.63 g):cholesterol (mw = 386.6 g) = 1:1 (molar ratio) with purified water (Milli-Q Plus, Millipore). The mixture was then incubated at 65°C for an hour, and then sonicated in a 65°C water-bath for 30 min. The sonicated mixture was then subjected to membrane protrusion (mini-Extruder, Avanti Polar Lipid, Ltd.) with a 200-nm pore size membrane (Avanti Polar Lipid, Ltd.), being extruded 20 times to form pre-liposome. The pre-liposome was then subjected to protrusion with a 100-nm pore size membrane, being extruded another 15 times. The liposome size and size distribution were then determined by photon correlation spectroscopy (Zetasizer Nano-ZS90, Malvern Instruments Ltd., Malvern, Worcestershire, UK). The 10-µl liposomes were dispersed with 500 µL purified water in a low volume disposable cuvette. The particle size and size distribution were measured in terms of ZAve and polydispersity index (PDI), respectively.

RNA sequencing for transcriptome analysis

RNA-Seq libraries were prepared using the Agilent SureSelect RNA Library Kit and were sequenced using Illumina Hiseq4000 150-nt PairedEnd to produce the reads (~25 million reads per sample). The reads were aligned with TopHat 2.0.13 to GRCh38 with default parameters and then were assembled by Cufflink 2.2.1 using Ensembl v79 annotations. Gene expression was measured in fragments per kb of exon per million fragments mapped (FPKM). For the differentially expressed genes regulated with LDLR, we performed functional enrichment analysis, as described in our previous studies (Cheng et al. 2014, Chung et al. 2015), to interpret their biological functions.

Statistics

The Student t-test or the chi-square analysis was used to identify significant differences between groups or categorical variables. A P value less than 0.05 was considered significant. All data are reported as the mean ± S.E.M.

Results

LDLR expression is a major determinant of platinum chemosensitivity

Given our hypothesis that the entry of tumor macroenvironmental lipid through the lipoprotein/receptor-route might contribute to the development of EOC subtypes, we examined the expression of LDLR-route-related genes (Chang et al. 2017) in EOC patients. The LDL/R-route components include LDLR (the entrance gate for LDL) and LPL (lipoprotein lipase; which is responsible for unloading lipid from LDL). As indicated by the results shown in Fig. 1A, B and C, LDLR and LPL were expressed throughout the various EOC subtypes. However, predominantly weak LDLR staining was seen in serous and mucinous EOC patients, whereas predominantly strong LDLR staining was seen in endometrioid and clear-cell EOC patients. Compared to that for LDLR, the expression pattern of LPL was more consistent across the various subtypes of EOC.

In order to validate the roles of LDLR in the responses of EOC cells to chemotherapy, we examined the LDLR expressions of EOC cells from serous (OVCAR3, SKOV3), endometrioid (MDAH-2774, TOV-112D), and clear-cell (TOV-21G, ES2) EOCs (Fig. 1D). We found that the endometrioid and clear-cell EOCs had abundant LDLR. Meanwhile, serous EOC was the subtype most sensitive to cisplatin, with a low IC50 value compared to those for the endometrioid and clear-cell EOCs (Fig. 1E). A similar phenomena was observed in the colony formation assay results (Fig. 1F).

In order to examine the impact of LDLR expressions on cisplatin responsiveness in vitro, shRNA targeting LDLR expression was introduced in endometrioid and clear-cell EOCs, while stably transfected LDLR cDNA was introduced in serous EOC (Fig. 2A). We found that the LDLR cDNA could reduce, while the shRNA could enhance cisplatin
colony-suppressing capacity (Fig. 2B and C) and cytotoxic efficacy (Fig. 2D and E). Furthermore, the IC 50 value of cisplatin could be significantly altered in TOV-112D (from 427 ± 27.4 to 111 ± 32.6 µM), TOV-21G (from 178 ± 72.7 to 99 ± 33.7 µM), MDAH-2774 (from 125 ± 26.6 to 63 ± 0.5 µM), and SKOV3 (from 8.9 ± 1.06 to 297.2 ± 15.43 µM) cells. In brief, the results shown in Figs 1 and 2 demonstrated that LDLR expression could potentially serve as a valuable biomarker for cisplatin regimens.

**Trans-Omics bioinformatics approach illustrated potential mechanism**

Considering the nature of LDLR for lipid importing, we performed lipidome analysis of three subtypes of EOC cells using shotgun lipid profiling technology. The lipid profiles were differentially expressed in six EOC cell lines, including with specific preferences (Fig. 3A). The ether-linked phospholipids were highly expressed in serous and endometrioid EOC cells. On the other hand, the storage lipids (DAG and TAG, which exist in endoplasmic reticulum and LD) were predominantly expressed in endometrioid and clear-cell EOC cells.

In order to determine the molecular mechanism of LDLR-mediated cisplatin insensitivity, we subjected two types of EOC cells (MDAH-2774 and TOV-21G; control vs knockdown LDLR) to transcriptome with RNA next-generation sequencing technology (RNAseq) and lipidome with shotgun LC-MS/MS profiling. The transcriptome and lipidome were then subjected to Trans-Omics analyses. Using integrative platform of transcriptome and Lipidome (unpublished works), we were able to associate genes function with GO terms and KEGG lipid metabolism map to interpret the relationship of gene and lipid expression acquired in the experiments.

The lipidome analysis revealed that lysophosphatidylcholine (LPC; upregulated; red-color labeled) and ether-linked phosphatidylethanolamine (PE O-; downregulated; black-bold labeled) were significantly altered by LDLR knockdown (Fig. 3C). Under the same experimental design, we found that 1404 genes were consistently altered (Fig. 3D). Through gene ontology-based annotation and functional enrichment analysis, we were then able to determine the top ten enriched pathways in terms of molecular function, which are listed in Fig. 3E. It was surprising to find that the lipid metabolism-related pathways were not prioritized in the LDLR-knockdown list, whereas transmembrane receptor activity and transcription factor-binding activity were.

The lysophatidyl lipids and phosphatidyl lipids, which are involved in the Lands cycle (converting by LPCAT1/2/3 and PLA2) (Moessinger et al. 2014), differ in
lyso-groups, so the trans-omics analysis strategies were designed as follows: 1st, transmembrane receptor activity genes were revealed by transcriptome analyses; 2nd, lyso-phospholipid-related phospholipase genes were revealed in transcriptome analyses; 3rd, phospholipase expression was negatively correlated with LDLR expression (for knockdown experimental design); and 4th, phospholipase genes significantly correlated to cancer survival were revealed in TCGA (The Cancer Genome Atlas) database.

The analysis of the Fig. 3D results shown in Fig. 3F indicated that there were no genes identified via all four strategies. Interestingly, however, there was one gene identified via strategies 1, 2, and 3 (SNCB, synuclein B; upregulated); one identified via strategies 1 and 3 (FAM83B, family with sequence similarity 83 member B; downregulated); and one identified via strategies 1 and 4 (VEGFR; downregulated). There were also two genes identified via strategies 1, 2, and 4 (FGFR1 and FGFR3; upregulated). Those analysis results suggested an indirect regulation of phospholipase (FAM83B or SNCB), LDLR, and RTK (VEGFR or FGFRs) expressions in EOC cells.

**LDLR → FAM83B → FGFR3 regulatory axis in platinum sensitivity**

In order to verify the correlations of regulation among gene expressions, we examined the LDLR-knockdown effect on SNCB, FAM83B, and RTKs. The results were consistent with the transcriptome (Fig. 3F) in that FGFR1–3 were downregulated (Fig. 4A). Meanwhile, knockdown LDLR also downregulated both protein (Fig. 4B, upper panel) and mRNA (Fig. 4B, lower panels) of FAM83B in EOC cells. In order to characterize the relationship of LDLR vs FAM83B vs FGFRs, we knocked down FAM83B in EOC cells. We found knockdown FAM83B did not alter LDLR, but downregulated FGFR1–3 in mRNA level (Fig. 4C). This indicating FAM83B is the downstream of LDLR, and upstream of FGFRs. As we verify the consequences of LDLR→FAM83B→FGFRs axis, we examined the FGFR-related downstream signals. We found pFAK, total FAK, and total MEK abundance were significantly decreased (Fig. 4D). At last, we used TCGA database with DriverDB. v2 platform to analyze the association of FAM83B in EOC survival, and we found it to be a negative risk factor for patients (Fig. 4E).

We then tested whether the LDLR-mediated FAM83B→FGFRs regulatory axis affects platinum sensitivity in the context of an LPC-associated event. We compared the cytotoxic effects of direct treatment with cisplatin, treatment with liposome-encapsulated cisplatin (DOTOP-liposome; 1,2-dioleoyl-3-trimethylammoniumpropane), and treatment with LPC-liposome-encapsulated cisplatin on the insensitive MDAH-2774 cells. Only 20 µM of cisplatin encapsulated in LPC-liposome exhibited excellent cytotoxic efficacy (Fig. 5A; lane 1 vs 5 vs 6). Meanwhile, neither LPC-cisplatin co-treatment (non-liposome) (Fig. 5A; lane 6 vs 7) nor DOTAP-liposome-cisplatin (Fig. 5A; lane 3 vs 4) showed such efficacy. When the effect of FAM83B knockdown on cisplatin cytotoxicity was tested, it was found that two clones of FAM83B knockdown could significantly facilitate cisplatin cytotoxic activity in MDAH-2774 cells.
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Figure 3
Transcriptome vs lipidome trans-omics approaches for potential molecular regulations. (A) Heat-map lipid profiling of six types of EOC cells, including serous (SKOV3, OVCAR3), endometrioid (MDAH-2774, TOV-112D), and clear-cell (ES2, TOV-21G) lines. The spectrum from blue to red indicates the variation of lipid species among the cells. (B) Lipidome analyses of lipid species comparing shLuc vs. shLDLR in MDAH-2774 and TOV-21G cells. The Y-axis shows the lipid species change P value, where the threshold > −log_{10}(1.4) indicates significant alteration (red-colored). LPC was increased and PE O- was decreased when the LDLR was knocked down. (C) Replicated transcriptome analysis by RNAseq was performed to compare shLuc vs shLDLR in MDAH-2774 and TOV-21G cells. The overlapped transcriptome showed that 1404 genes were consistently altered. (D) Pathway enrichment analysis by GO-term followed with GSEA analysis. The top 10 enriched pathways were ranked from highest to lowest P value (−log_{10}). (E) Trans-omics analyses of transcriptome and lipidome profiles. The four selection criteria were implemented in the analyses. Strategy 1: transmembrane receptor activity genes were revealed by transcriptome analyses; Strategy 2: lyso-phospholipid-related phospholipase genes were revealed in transcriptome analyses; Strategy 3: phospholipase expression was negatively correlated with LDLR expression (for knockdown experimental design); Strategy 4: phospholipase genes significantly correlated to cancer survival were revealed in TCGA database. The results came out SNCB (1 ∩ 2 ∩ 3), FAM83B (1 ∩ 3), FGFR1/3 (2 ∩ 4), and FGFR1 (1 ∩ 4). The lipid profiling data were from two replicated experiments, and the gene expression and cytotoxic assay results were from at least three independent experiments.

In testing the LPC-liposome effect on gene expressions, meanwhile, we found that both FAM83B (Fig. 5C) and FGFRs (Fig. 5D) were upregulated in comparison with DOTOP liposome. Except for the gene regulation in cultured EOC cells, we also examined LDLR expression in association with FGFRs in TCGA database. We found that LDLR is moderately reverse associated with FGFR1 and FGFR3 in patients (data not show), which was consistent with the data we found for EOC cells.

To further test the effects of LDLR expression on cisplatin responsiveness in vivo, we performed the xenograft tumor with cisplatin treatment procedure (see ‘Methods’ section) in mice bearing MDAH-2774, TOV-21G, and SKOV3 EOC cells. We found that the tumor with shLuc-infected MDAH-2774 cells could only exerted a minor degree of tumor suppression by Cisplatin treatments; however, tumors with LDLR-knockdown tumors caused by shLDLR were dramatically ameliorated by the same treatment procedure (Fig. 6A). In contrast, the tumors with pLenti-infected SKOV3 cells could be inhibited by the cisplatin treatment procedure, and the effects of treatment were comparable in tumors in which LDLR cDNA was stably expressed (Fig. 6B). To further confirm the effects of LDLR expression on cisplatin responsiveness, we performed the same procedure in tumors with shLuc-infected TOV-21G cells and found...
very little response. However, the tumors with shLDLR-infected TOV-21G cells could be almost abolished with the treatment procedure (Fig. 6C).

In brief, the results of the mechanistic dissection shown in Figs 4 and 5 revealed an LDLR→LPC→FAM83B→FGFRs→platinum insensitivity axis in EOC, while the results shown in Fig. 6 supported the conclusion that the expression of LDLR in EOCs determined their cisplatin responsiveness.

**Discussion**

In this study, we found that the differential expression of the LDLR-route in EOCs, as revealed in histological variations, determines the platinum-based therapy insensitivity of those EOCs. The mechanism underlying this could involve the LDLR→LPC→FAM83B→FGFRs regulatory axis. The mechanism of LDLR in platinum chemosensitivity can be explained in several steps as the illustrative scheme shown in Fig. 6D. First, the circulating LDL could affect cancer cells; second, the LDLR engulfs LDL in association with LDLRAP (LDLR Associate Protein); third, the engulfed LDL could then unload lipids by LPL.
The imported lipidome could then decrease LPC amount. Moreover, the LPC amount could trigger the FAM83B expression and the consequence FGFRs expression; fourth, alternation in LPC, FAM83B and FGFRs expressions could lead to cisplatin insensitivity. The potential impacts of our findings are discussed below:

Translational value of LDLR as a biomarker for chemotherapy responsiveness

The effects of LDL/R route and related lipidome expression on chemosensitivity have previously been hypothesized but had not previously been verified (Huang et al. 2016). With regard to platinum-based therapy, it was previously shown that the latency from initial adjuvant platinum therapy to resistance is around 90–250 days in lung adenocarcinoma (Wu et al. 2015). It has been speculated that the mechanism underlying this resistance could involve cholesterol-induced ABCG2 expression, with the ABCG2 being recognized as a channel for pumping out lipophilic wastes, for example, platinum chemoagents. Interestingly, LD function is related to lipid transport through ABC proteins (Baldan et al. 2009, Gulati et al. 2015). Relatedly, a recent study using transcriptome and metabolomics analyses in NCI-60 cell lines found that lipoprotein uptake is one of the hallmark events in platinum sensitivity (Cavill et al. 2011). As for other chemoagents, other evidence has shown that cholesterol uptake (potentially via LDLR) could be an important event for developing gemcitabine resistance in cases of pancreatic cancer, which has caused such uptake to be considered an excellent therapeutic target (Guillaumond et al. 2015). In this study, we discovered an LDLR→LPC→FAM83B→FGFRs axis for platinum sensitivity. Taking advantage of this discovery, we then implemented the use of LPC-liposome-encapsulated cisplatin as a new chemoagent (Fig. 5A). In terms of translating to clinical usage, further delicate technical and efficacy improvement of...
LPC-liposome-cisplatin in multiple cancer types and in vivo delivery should be conducted.

Targeting LDLR as part of future therapies for EOCs

Proprotein convertase subtilisin/kexin type 9 (PCSK9) is a soluble member of the mammalian proprotein convertase family of secretory serine endopeptidases (Seidah et al. 2003). Circulating PCSK9 is known to degrade LDLR (Lagace et al. 2006) via binding on the EGF-like repeat A site of LDLR, which causes the recycling activity to be reduced and, thus, causes the degradation of LDLR (Zhang et al. 2007). The mechanism-of-action of PCSK-9 inhibitors (such as alirocumab and evolocumab) exerts anti-degrading liver LDLR activity; therefore, these PCSK-9 inhibitors enhance LDLR recycling to reduce circulating LDL levels. In this study, we discovered that the LDL/R route reduces platinum efficacy. Therefore, one of the strategies for targeting the LDL/R-route consists of reducing systemic LDL levels. Introducing a PCSK-9 inhibitor in the newly developed LPC-liposome-cisplatin for this purpose exhibited great potential. On the other hand, an LDL/R-route-targeting strategy could also include degraded cancer LDLR. One in vitro study showed that PCSK-9 could degrade LDLR by interacting with gyspican-3 in liver cancer cells (Ly et al. 2016). The development of PCSK-9 peptide for degrading LDLR (Lagace 2014) is also a hypothetical strategy for platinum-based therapy.

In conclusion, this study discovered that LDLR expression is a major determinant for platinum chemosensitivity. Furthermore, the novel LDLR→LPC→FAM83B→FGFRs regulatory axis revealed by the trans-omics analyses conducted in this investigation might explain platinum chemosensitivity discrepancies.

Supplementary materials
This is linked to the online version of the paper at https://doi.org/10.1530/ERC-19-0095.

Declaration of interest
The authors declare that there is no conflict of interest that could be perceived as prejudicing the impartiality of the research reported.

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Author contribution statement
W C Chang collected the clinical samples, conceptualized the study, and drafted the manuscript. H C Wang performed the animal experiment and drafted the methodology section, while W C Cheng performed the bioinformatics analysis. J C Yang performed HPLC for the platinum measurements and assisted with the in vitro experiment. W M Chung and Y P Ho helped with the immunohistochemistry study. Y C Hung supported the study, analyzed the data, and edited the manuscript. W L Ma conceptualized the study, coordinated the research project, supported the project, and edited/approved the final version of the manuscript.

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