RNA interference (RNAi) screening approach identifies agents that enhance paclitaxel activity in breast cancer cells

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Abstract

Introduction: Paclitaxel is a widely used drug in the treatment of patients with locally advanced and metastatic breast cancer. However, only a small portion of patients have a complete response to paclitaxel-based chemotherapy, and many patients are resistant. Strategies that increase sensitivity and limit resistance to paclitaxel would be of clinical use, especially for patients with triple-negative breast cancer (TNBC).

Methods: We generated a gene set from overlay of the druggable genome and a collection of genomically deregulated gene transcripts in breast cancer. We used loss-of-function RNA interference (RNAi) to identify gene products in this set that, when targeted, increase paclitaxel sensitivity. Pharmacological agents that targeted the top scoring hits/genes from our RNAi screens were used in combination with paclitaxel, and the effects on the growth of various breast cancer cell lines were determined.

Results: RNAi screens performed herein were validated by identification of genes in pathways that, when previously targeted, enhanced paclitaxel sensitivity in the pre-clinical and clinical settings. When chemical inhibitors, CCT007093 and mithramycin, against two top hits in our screen, PPMID and SP1, respectively, were used in combination with paclitaxel, we observed synergistic growth inhibition in both 2D and 3D breast cancer cell cultures. The transforming growth factor beta (TGFβ) receptor inhibitor, LY2109761, that targets the signaling pathway of another top scoring hit, TGFβ1, was synergistic with paclitaxel when used in combination on select breast cancer cell lines grown in 3D culture. We also determined the relative paclitaxel sensitivity of 22 TNBC cell lines and identified 18 drug-sensitive and four drug-resistant cell lines. Of significance, we found that both CCT007093 and mithramycin, when used in combination with paclitaxel, resulted in synergistic inhibition of the four paclitaxel-resistant TNBC cell lines.

Conclusions: RNAi screening can identify druggable targets and novel drug combinations that can sensitize breast cancer cells to paclitaxel. This genomic-based approach can be applied to a multitude of tumor-derived cell lines and drug treatments to generate requisite pre-clinical data for new drug combination therapies to pursue in clinical investigations.

Introduction

Chemotherapy regimens containing taxanes, including docetaxel and paclitaxel, have well-established benefits in breast cancer [1,2]. Despite improvement in the response rates with use of taxane-based drug combinations versus single agent taxanes, most patients do not have a complete response to treatment [3-6]. A partial response or resistance to paclitaxel is a major limiting factor in the successful treatment of breast cancer. Improving taxane-based chemotherapy regimens through novel drug combinations is therefore of clinical interest. Patients with tumors that lack expression of estrogen receptor (ER), progesterone receptor (PR), and HER2 amplification (triple-negative breast cancer, TNBC) are not candidates for currently available FDA-approved, targeted therapies.
More efficacious combination chemotherapy is needed for these patients.

Due to its extensive use in breast cancer and other tumor types and the frequency of acquired resistance, mechanisms of taxane resistance have been investigated [7-9]. Some mechanisms identified to date include mutations of the β-tubulin gene [10,11], expression of the tubulin binding protein tau [12], expression of ER [13,14], HER2 [15,16], BRCA1 [17,18], and p-glycoprotein/MDR1 [19-21], among others [8,9]. Genomic studies have also been used for predicting response to both paclitaxel and related compound docetaxel [3,5,6,22,23], but few if any genes amongst these studies overlap or have been confirmed as reliable markers or predictors of response. Despite these studies, novel therapeutic combinations with paclitaxel are being tested in clinical trials, especially in patients with advanced disease or those without clinically proven therapeutic targets such as TNBC [24-26]. Identification of gene products that when pharmacologically inhibited enhance paclitaxel sensitivity may lead to improved response rates and reduced resistance.

The advent of RNA interference (RNAi) for gene silencing allows for systematic gene and/or pathway analysis in tumor cells and an ability to uncover novel gene functions and pathways that cannot always be identified by ectopic gene expression. Several RNAi studies performed in human tumor cell lines using synthetic small interfering RNAs (siRNAs) or vector-based short hairpin RNAs (shRNAs) targeting defined gene families or genome-wide libraries have identified modulators of drug sensitivity [27-33]. These studies have unveiled novel pathways and molecules for therapeutic targeting in various tumor types and there is a great need to translate this information for clinical utility.

Genomic tumor profiling has provided us with important insights to mechanisms of tumorigenesis and translational data for clinical advances. Relative to some cancer types, there is tremendous genomic information available for breast cancers, which includes tumor DNA copy number [34-38], DNA sequence and mutations [39-44], gene expression and protein profiles [45,46], as well as epigenetics [47,48] and microRNAs [49,50]. In the current study, we performed genetic loss-of-function RNAi screens to identify druggable targets involved in paclitaxel sensitivity. In our screens, we used a gene set that is comprised of the overlay of a druggable genome library with a set of genes considered to be deregulated in breast cancer (from genomic studies of human breast cancers and cell lines [37,38]). Specific pharmacological inhibitors of the top scoring hits from our screens were used in combination with paclitaxel and the ability of the chemicals to enhance the growth inhibitory activity of paclitaxel on breast tumor-derived cell lines was analyzed. We further tested these novel paclitaxel drug combinations on four paclitaxel-resistant TNBC cell lines and for select inhibitors showed synergistic drug activity. New findings presented in this study show the feasibility of loss-of-function screening to provide biological relevance for genomic discoveries and to identify drug combinations to improve current taxane-based drug treatments in preclinical models for breast cancer.

Materials and methods

Reagents and resources
Paclitaxel, CCT007093, and mithramycin A (Sigma-Aldrich, St. Louis, MO, USA) were prepared in DMSO at a stock concentration of 0.1 mM, 5 mM, and 0.9 mM, respectively. LY2109761 was kindly provided by Jonathan Yingling, Lilly Research Laboratories, Indianapolis, IN, USA and prepared in DMSO at 10 mM stock concentration. The panel of candidate genes used in the shRNA screen was generated from overlay of a list of 1,778 genomically deregulated gene transcripts whose levels significantly correlated with genome copy number in breast cancer [37,38] and a druggable genome list compiled from two sources (Open Biosystems, Huntsville, AL, USA and Qiagen, Valencia, CA, USA). Pharmacological agents were identified using several drug databases including DrugBank, Therapeutic Target Database, Comparative Toxicogenomics Database, and Ingenuity Pathway Analysis.

Cell culture
HeLa and MCF-7 cells were purchased from American Tissue Cell Culture (ATCC, Manassas, VA, USA) and cultured in Dulbecco’s modified Eagle’s medium (DMEM, Invitrogen, Carlsbad, CA, USA) supplemented with 10% fetal bovine serum, and 1% penicillin-streptomycin. All TNBC cell lines were purchased from ATCC or Deutsche Sammlung von Mikroorganismen und Zellkulturen GmbH (DSMZ, Braunschweig, Germany) and cultured as described (Additional File 1). All cells were cultured at 37°C with 5% CO2 and tested routinely for mycoplasma, using the MycoAlert Detection Kit (Cambrex, Rockland, ME, USA).

shRNA and siRNA screens
HeLa cells were plated at 20,000 cells per well (96-well plate) and 24 h later transfected with a subset of the human genome pGIPZ shRNAmir plasmid library (n = 1,078) (Open Biosystems), as provided by the Functional Genomics Shared Resource at Vanderbilt University in a one clone per well format. The next day, cells were split 1:6 into 96-well plates, allowed to attach overnight, and three plates were treated with vehicle control (DMSO) and three were treated with 5 nM paclitaxel for 24 h. Cells were washed, replaced with fresh media and incubated for an additional 72 to 96 h. Alamar Blue (Invitrogen), a...
dye used to detect metabolic activity in cells, was used to assay for cell viability and to identify genes that alter paclitaxel sensitivity. To identify gene targets that promote paclitaxel sensitivity or resistance, we generated a sensitivity index (SI) score for each shRNA obtained from replicate experiments after drug treatment [32]. The SI score accounts for both the individual effect of shRNAs and the effect of drug on cell viability (see next section for description of the statistical methodology). Data from each plate were normalized to non-silencing (NS) shRNA controls that do not target any human gene, to account for plate-to-plate variability and to control for the effects of shRNA transfection. For the siRNA screen, two independent siRNAs were designed for each gene and randomly distributed in a 96-well plate. MDA-MB-231 and MDA-MB-468 cells were reverse-transfected with siRNAs complexed with lipid reagent for 48 h and subsequently split into four replicate plates. Cells were treated and measured for viability in a similar fashion as above. Transfections (that is, experiments) were performed in triplicate to allow for assessment of variation of expression data in statistical analysis.

Statistical analysis
Median centered global normalization was performed across all shRNA and siRNA plates by using the NS controls in each plate. The SI score was calculated for each of the shRNAs and siRNAs by estimating the difference between the expected and observed combined effects of shRNAs or siRNAs and paclitaxel on cell viability, as previously described [32]. The SI scores range from -1 to 1. Positive SI scores indicate sensitizing effects and negative SI scores indicate antagonizing effects.

A bootstrap algorithm was used to estimate the variability of the mean SI level for each gene with > 3 shRNAs by randomly sampled from all shRNAs of that gene with replacement. The corresponding 95% percentile bootstrap confidence interval was calculated for each gene. Genes were taken as hits if they had a mean bootstrap in the upper quartile cutoff SI > 0.078 and the lower bound of 95% confidence interval > 0. The results of a small simulation study we carried out show that the bootstrap distribution from a very small number of shRNAs (≤3 per gene) is not reliable. Therefore, the mean SI value was calculated for the genes with ≤3 shRNAs. A more stringent cutoff (SI > 0.15) was used for hit selection among these genes. For the siRNA screen, the SI value was calculated by averaging the two siRNAs for each gene after normalization and the top hits for each cell line were selected based on the SI value of the averaged data. Correlation between experiments was estimated using Pearson’s correlation coefficient. Statistical analysis was performed using R software (version 2.10.1).

Cell growth and viability assays
For cell growth assays cells were seeded at 5 × 10^5 cells per well of a six-well plate. The next day cells were treated with 5 μM CCT007093 or 10 nM mithramycin, ± 3 nM paclitaxel, or vehicle control (DMSO). After three days cells were collected, washed, and counted using a Coulter Counter (Beckman-Coulter, Brea, CA, USA). Cell number was plotted as a percent of cells relative to vehicle control. Cell viability assays were performed by seeding 3,000 to 8,000 cells per well of a 96-well plate. The next day, growth media was replaced with treatment media containing vehicle-DMSO or paclitaxel that was serial diluted by half-log concentrations ranging from 0.3 to 30 nM. After three days of incubation with the drug, cell viability was measured using the Alamar Blue assay (Invitrogen). Cell viability for each drug concentration was compared to vehicle-treated control. Four replicate wells from three independent experiments of each drug concentration were used to generate median-effect plots to calculate the IC_{50} (concentration for 50% growth inhibition) concentrations for each cell line using CalcuSyn Software (Biosoft, Cambridge, United Kingdom). IC_{50} values for each cell line are represented with standard error.

Mammosphere cultures
For three-dimensional (3D) mammosphere cultures, cells were seeded on growth factor-reduced Matrigel (BD Biosciences, San Jose, CA, USA) in chamber slides as previously described [51,52]. CCT007093, mithramycin, and LY2109761 ± paclitaxel were added to medium 24 h after cell seeding and medium was replaced every three days. Mammospheres were detached from Matrigel with dispase enzyme (BD Biosciences), trypsinized into single cell suspensions, and cell number was determined using a hemocytometer. The number of viable cells was plotted as a percent of cells relative to vehicle control.

Drug synergy analysis
Paclitaxel was combined with each of the different agents at a fixed ratio (1:1) of the individual IC_{50} concentrations of each drug. Drug combinations were then serial diluted (1:2) and represented as IC_{50} IC_{25} and IC_{12,5} concentrations, as the additive effects of both drugs. Statistical analysis of drug synergy was evaluated from the results of the Alamar Blue assays and calculated using the Chou-Talaly method [53] and CalcuSyn Software (Biosoft). To determine synergy between two drugs, the software uses a median-effect method that determines if the drug combination produces greater effects together than expected from the summation of their individual effects. The combination index (CI) values are calculated for the different dose-effect plots (for each of the serial dilutions) based
on the parameters derived from the median-effect plots of the individual drugs or drug combinations at the fixed ratios. The CI was calculated based on the assumption of mutually nonexclusive drug interactions. CI values significantly > 1 are antagonistic, not significantly different than 1 are additive, and values < 1 are synergistic. Two-sided statistical tests were used to determine if the mean CI values resulting from three independent experiments at multiple effect levels were statistically significantly different from a CI = 1.

**Results**

**RNAi screening for genes that sensitize cells to paclitaxel**

To identify druggable gene targets that could enhance paclitaxel activity in breast cancer cells, we performed an shRNA screen. We selected a subset of genes based on a comprehensive genomic study of 145 primary human breast tumors and 51 breast cancer cell lines in which 1,778 gene transcripts were identified whose levels significantly correlated with genome copy number and are deemed genomically deregulated in breast cancer [37]. Most of the alterations present in primary tumors were retained in the cell lines [37]. The 1,778 genomically deregulated genes were overlaid with a druggable gene list (compiled from two sources, Open Biosystems and Qiagen), with the expectation that for select genes identified in the shRNA screen, an agent may already exist that could be analyzed in preclinical models for synergistic activity with paclitaxel. The overlay of the gene lists yielded 428 genes (Figure 1A). From a whole-genome vector-based shRNAmir library, we generated a sub-library consisting of 1,078 shRNAs targeting the 428 genes, with 1 to 11 shRNAs per gene. Since the transfection efficiency of plasmid-based vectors in most breast cancer cell lines is < 10%, we used a highly transfactable cell line, HeLa, for our primary screen with the assumption that genes/pathways related to paclitaxel sensitivity are conserved across cancer cell lines. Positive hits from the first screen in HeLa cells were validated in secondary screens using two triple-negative breast cancer (TNBC) cell lines as described below.

shRNAs for each gene in our sub-library were independently transfected into HeLa cells in a 96-well-plate format and cells were split 24 h after transfection into six replicate plates. After 48 h, half of the plates (n = 3) received an IC_{50} concentration of paclitaxel (5 nM) and half received vehicle (DMSO) treatment. In order to detect significant differences in drug sensitivity in the assay, we allowed time for multiple cell divisions. After four days of drug treatment, cell viability was measured using an Alamar Blue assay to identify genes that alter paclitaxel sensitivity (effect of shRNA and drug). Comparison of the mean viability values of three replicates for each shRNA from the two individual screens revealed high reproducibility (r = 0.89, Pearson’s correlation coefficient) (Figure 1B). We combined the results from the duplicate screens in the final analyses.

To account for plate-to-plate variability, we normalized across all the plates using non-silencing (NS) control shRNAs that were present in each plate. To identify genes that when targeted promote paclitaxel sensitivity or resistance; we generated a sensitivity index (SI) score for each shRNA obtained from replicate experiments after drug treatment, as previously described (Figure 1C) [32]. The SI score accounts for the individual effect of shRNAs and the effect of drug on cell viability. A positive SI score is a measure of sensitivity and a negative SI score is indicative of resistance to paclitaxel treatment. In this study, we chose gene targets that are amplified/overexpressed in breast and that increase paclitaxel sensitivity (+SI value), as these are more likely to be better targets for pharmacological inhibition.

For selection of hits from our primary shRNA screen, we used a bootstrap algorithm to identify gene targets that had > 3 shRNAs based on the mean SI > 0.078 (upper quartile) and the corresponding 95% confidence interval (Table 1). These criteria allowed for high-confidence hits to be selected. As the number of positive scoring (+SI) shRNAs for each gene increased, our confidence for these genes increased, as these are unlikely due to false-positives or off-target effects of individual shRNAs. However, since this method biased our hit selection for those genes that had more shRNAs in our sub-library, we selected additional hits represented by genes that had ≤3 shRNAs but with a much more stringent cutoff of mean SI value > 0.150 (Table 1). FRAP1 (mTOR) (mean SI = 0.212; Table 1) was previously identified through an RNAi screen as a target of paclitaxel sensitivity, and was used in our screen as a positive control in each plate [30]. CASP3 shRNA (mean SI = -0.042) was used as a negative control in each plate as we found that this gene, when downregulated, induces paclitaxel resistance (Table 1). Three of the four shRNAs that target EGFR were highly sensitive to paclitaxel activity (mean SI = 0.136, Table 1). EGFR is a known target of paclitaxel sensitivity as erlotinib, an EGFR inhibitor, increases paclitaxel activity in vivo [54-57]. Additionally, TUBG1, tubulin gamma-1, a component of the γ-tubulin ring complex (γ-TuRC), involved in mitotic spindle formation, enhanced paclitaxel sensitivity (mean SI value = 0.152, Table 1). γ-TuRC has previously been shown to enhance paclitaxel sensitivity, in vitro [33]. These data collectively validated our primary shRNA screening approach.

To determine if the results of the shRNA screen were reproducible in breast cancer cells, we validated the top 36 high-confidence hits (genes) from the shRNA screen that were amplified/overexpressed in breast cancer and had positive SI values (Table 1). Some of the genes...
Figure 1 shRNA screen to identify paclitaxel sensitizers. A. The overlay of druggable genome libraries (Qiagen and Open Biosystems) and genes deregulated in breast cancer resulted in 428 candidate druggable genes. B. Reproducibility of shRNA screen by correlation of the effect of shRNAs on cell growth compared to non-silencing shRNA in vehicle-treated plates of two replicate experiments. Spearman correlation coefficient, r = 0.89. C. Each shRNA was scored for the level of paclitaxel sensitivity using the sensitivity index (SI) as described in Materials and Methods. The SI score ranges from -1 to 1. Positive significant SI scores indicate sensitization and negative significant SI scores indicate antagonism. The scatter plot of all shRNAs is shown in rank order. The dashed lines indicate the relative threshold of significant drug sensitivity.
selected are targets of agents that have not been tested for efficacy in combination with paclitaxel in the preclinical setting and are of biological relevance and interest (for example, transforming growth factor beta (TGFβ) signaling). Two independent siRNA oligos were designed for each of the 36 genes selected and reverse-transfected into two TNBC cell lines, MDA-MB-231 and MDA-MB-468. Duplicate experiments were performed and resulted in high reproducibility (correlation coefficients approximately 0.70 to 0.80, data not shown).

We averaged the SI value for the two siRNAs from duplicate experiments for each gene and the top hits for each cell line were selected for further analysis (Table 2). Four genes, PPM1D, CENPF, BCL2L1, and FRAP1 were sensitizers of paclitaxel in both cell lines (bold, Table 2). Since paclitaxel efficacy is dependent on mitotic activity (that is, cell cycle transit into M-phase), we postulated that siRNAs that decreased cell viability > 30% in untreated plates were unlikely candidates for enhancing paclitaxel activity as cell cycle slowing or arrest limits the efficacy of paclitaxel. However, we did note the effect that some siRNAs had on breast cancer cell viability in untreated plates as the targeted gene may be of potential interest for further investigation for breast cancers that do not have targeted therapy, such as TNBC. For example, IGF1 siRNA in MDA-MB-468 cells led to a 60% reduction in viability compared to NS siRNA control (data not shown). However, we did not observe significant sensitivity to paclitaxel (SI = -0.031) for IGF1 siRNAs in these cells, likely due to the large loss of cell viability prior to paclitaxel treatment.

To ensure that drug sensitivity correlated with relative decreases in gene expression and to eliminate any possible off-target effects from shRNAs and siRNAs, we used Dharmacon ON-TARGETplus individual and pooled siRNAs as a third independent RNAi approach on select positive hits and our results with PPMID are shown as an example. ON-TARGETplus siRNAs for a top hit, PPMID, were transfected in two breast cancer cell lines, MCF-7 and MDA-MB-468. PPMID knockdown was measured at 48 h after transfection by quantitative real-time PCR. Three of the four individual and the pooled ON-TAR-

Table 1: Paclitaxel sensitivity index for indicated genes from shRNA screen

| Gene | shRNAs | Mean SI | 95% CI | Genomic dereg* | Gene | shRNAs | Mean SI | Genomic dereg* |
|------|--------|---------|--------|---------------|------|--------|---------|---------------|
| YWHAZ | 6      | 0.193   | 0.154 to 0.242 | amp/OE | PCK1  | 1      | 0.461   | amp/OE        |
| RP56KB1 | 4     | 0.186   | 0.135 to 0.242 | amp/OE | SREBF2 | 1      | 0.391   | amp/OE        |
| COG2  | 5      | 0.186   | 0.110 to 0.265  | amp/OE | SRC    | 1      | 0.276   | amp/OE        |
| PTK2  | 5      | 0.184   | 0.121 to 0.287  | amp/OE | FNTA   | 1      | 0.263   | amp/OE        |
| PPMID | 4      | 0.179   | 0.120 to 0.241  | amp/OE | BCL2L1 | 1      | 0.247   | amp/OE        |
| SKP1A | 6      | 0.166   | 0.075 to 0.259  | amp/OE | COMMD1 | 3      | 0.220   | amp/OE        |
| MARK1 | 5      | 0.157   | 0.006 to 0.345  | amp/OE | COG8   | 1      | 0.214   | amp/OE        |
| NFYB  | 4      | 0.148   | 0.085 to 0.186  | amp/OE | FRAP1  | 1      | 0.212   | NA            |
| RBBP4 | 4      | 0.139   | 0.080 to 0.198  | amp/OE | ERBB2  | 1      | 0.180   | amp/OE        |
| IL10  | 4      | 0.136   | 0.081 to 0.205  | amp/OE | IQGAP1 | 3      | 0.176   | amp/OE        |
| EGFR  | 4      | 0.136   | 0.087 to 0.378  | amp/OE | PHB    | 3      | 0.159   | amp/OE        |
| SP1   | 5      | 0.130   | 0.080 to 0.138  | amp/OE | NDUF56 | 1      | 0.159   | amp/OE        |
| STX16 | 4      | 0.107   | 0.053 to 0.175  | amp/OE | COG1   | 1      | 0.159   | amp/OE        |
| PTPN7 | 4      | 0.095   | 0.019 to 0.170  | amp/OE | PRPF4B | 3      | 0.156   | amp/OE        |
| SENP1 | 5      | 0.095   | 0.022 to 0.168  | amp/OE | FADD   | 3      | 0.156   | amp/OE        |
| CENPF | 4      | 0.085   | 0.030 to 0.140  | amp/OE | ERK1   | 1      | 0.154   | amp/OE        |
| IGF1  | 4      | 0.078   | 0.016 to 0.137  | amp/OE | TGBF1  | 1      | 0.153   | amp/OE        |
| CASP3 (control) | 5 | -0.042 | -0.061 to -0.021 | del/UE | IK8KB  | 2      | 0.151   | amp/OE        |

* as determined by DNA copy number and gene expression analysis by Neve et al.  
SI, sensitivity index; CI, confidence interval; amp/OE, amplification/overexpressed; del/UE, deletion/underexpressed. 
NA, not available.
GETplus siRNAs for PPM1D showed > 80% reduction in PPM1D mRNA levels in MCF-7 cells and > 60% reduction in MDA-MB-468 cells (Additional File 2). Importantly, knockdown of PPM1D was correlated with increased paclitaxel sensitivity over a range of paclitaxel doses in both cell lines (Additional File 2). The use of multiple shRNAs and validation with independent siRNAs limited the likelihood that the observed sensitivity was due to off-target effects.

### Candidate pharmacological inhibitors that enhance paclitaxel sensitivity

A primary goal of this study was to identify gene targets that are druggable, to which pharmacological agents have been developed, and that can be used in novel combinations with paclitaxel in preclinical studies. The list of top hits from the validation siRNA screen for both cell lines is shown in Table 2 with associated chemical agents identified using in silico drug databases (see Materials and Methods). In some cases, agents linked to genes in the list represent FDA-approved drugs, some of which have already been successfully used in combination with paclitaxel (for example, FRAP1; rapamycin [58,59], EGFR; erlotinib [54,60,61]). Gene targets with inhibitors known to enhance paclitaxel sensitivity both in preclinical [62-67] and clinical models [54,58,61,68,69] (noted in Table 2) were not studied further; however, their discovery validated our RNAi screening approach. We also did not pursue hits that had non-specific inhibitors and those that had no available agents despite being considered druggable (for example, MARK1); however, those gene targets still remain of interest. Since some hits are involved in intricate signaling pathways, there could be other drug targetable molecules within the same pathway, which could impact paclitaxel sensitivity. For example, a top hit in our screen, RPS6KB1, is downstream of mTOR and PI3K, two prominent signaling pathways in breast cancer with known direct inhibitors, rapamycin and LY294002, and that have been shown to sensitize cells to paclitaxel [59,70].

Three gene targets from our list were of particular interest. These genes encode proteins to which agents have been developed and thus we could test the compounds in combination with paclitaxel for biological effect. The first was PPM1D, a member of the PP2C family of serine/threonine protein phosphatases, and a known negative regulator of cell stress response pathways including those regulated by p53, CHEK1, and p38 MAP kinase [71]. PPM1D is amplified and overexpressed in breast cancers [72,73] and inhibition of its activity, through use of small molecules such as CCT007093, inhibits the growth of tumor cell lines that overexpress PPM1D [74,75]. The second gene target of interest was SP1, a constitutively expressed transcription factor that

| Gene   | MDA-MB-231 Mean SI | MDA-MB-468 Mean SI | Drug/chemical inhibitor                                                                 | Previous combination with paclitaxel | Pre-clinical | Clinical | Ref   |
|-------|------------------|------------------|-----------------------------------------------------------------------------------|-----------------------------------|-------------|----------|-------|
| PPM1D | 0.055            | 0.136            | thioxanthen-9-one; CCT007093; anti-estrogens*                                      | no                                | no          |          |       |
| CENPF | 0.049            | 0.113            | farnesyltransferase inhibitors*                                                   | yes                               | yes         | [62,68,69]|
| BCL2L1| 0.041            | 0.093            | ABT-737; AT-101                                                                  | yes                               | no          | [63,64]  |
| FRAP1 | 0.037            | 0.078            | rapamycin; RAD001                                                                | yes                               | yes         | [58,59]  |
| IGFR  | 0.038            | 0.085            | NVP-AEW541; 9-cis-retinoic acid*; raloxifene*                                    | no                                | no          |          |       |
| EGFR  | 0.154            |                   | erlotinib; gefitinib; cetuximab                                                  | yes                               | yes         | [54,60,61]|
| ERK1  | 0.148            |                   | ERK/MEK inhibitors                                                              | yes                               | no          | [65,66]  |
| RPS6KB1| 0.140           |                   | rapamycin*; RAD001                                                            | yes                               | yes         | [58,59]  |
| TGFBI | 0.121            |                   | LY2109761; LY2157299; SD-208                                                    | no                                | no          |          |       |
| SP1   | 0.085            |                   | mithramycin; arsenic trioxide*                                                  | no                                | no          |          |       |

*indirect inhibitors
SI, sensitivity index
Novel drug combinations with paclitaxel inhibit 3D growth of breast cancer cell lines

To determine the effect of the novel drug combinations on paclitaxel sensitivity in 3D cultures, we grew two cell lines, MDA-MB-468 and MCF-7, as mammospheres, a culture method that has been developed to analyze breast epithelial function, morphology, and invasiveness [51,52]. Paclitaxel treatment alone reduced mammosphere formation and overall cell number by 37% in MCF-7 and 36% in MDA-MB-468 cells (Figure 3A, B). CCT007093 treatment alone reduced MCF-7 mammospheres by 46% versus <1% reduction of MDA-MB-468 mammospheres, a line that does not have appreciable PPM1D expression levels. However, CCT007093 treatment enhanced paclitaxel sensitivity and reduced mammosphere cell number by 89% in MDA-MB-468 and 92% in MCF-7 cultures. Likewise, we observed a significant reduction in the number of cells in the mammospheres that formed with the combination of mithramycin and paclitaxel in both MDA-MB-468 and MCF-7 cells (92% and 86% reduction, respectively). Although we did not observe any appreciable drug synergy with the TGFβR inhibitor LY2109761 in 2D, monolayer cell culture, we did observe a significant effect in 3D cultures. When used in combination with paclitaxel, LY2109761 inhibited mammosphere formation and reduced cell number by 72% and 92% compared to control in MDA-MB-468 and MCF-7 cells, respectively; however, it had minimal effect on mammosphere cell growth when used as a single agent (<20% reduction).

CCT007093 and mithramycin synergistically enhance paclitaxel activity in paclitaxel-sensitive and -resistance TNBC cell lines

There are currently no targeted therapies for patients with TNBC. Frequently, patients with this type of breast cancer receive paclitaxel, due to its initial effects and higher response rates as compared to other chemotherapies [83]. However, not all patients have a complete response and those that are resistant or have residual disease after initial or secondary chemotherapy have a worse prognosis and outcome [83,84]. In addition, TNBC patients that initially respond to chemotherapy have a higher incidence and faster relapse compared to patients with non-TNBC [85]. Thus, improving the effect of initial paclitaxel treatment is an important goal in successfully treating patients with TNBC until more improved and/or targeted therapies are developed.

Along these lines, we determined the relative paclitaxel sensitivity of a panel of TNBC cell lines by determining the paclitaxel IC_{50} values for 22 TNBC cell lines (Figure 4A). The distribution of IC_{50} values across the panel led us to classify 18 cell lines as relatively paclitaxel-sensitive and four cell lines (CAL120, SW527, HDQPI, and MT3, which had relatively high IC_{50} values (>20 nM)) as relatively paclitaxel-resistant. We determined if the four resistant cell lines could be sensitized to paclitaxel using the novel drug combinations presented above and...
Figure 2 Novel drug combinations sensitize breast cancer cells to paclitaxel. **A**. MDA-MB-231, MDA-MB-468, and MCF-7 breast cancer cell lines were seeded in six-well plates and treated with vehicle (control), < IC_{50} concentrations of the putative PPM1D inhibitor, CCT007093 (CCT); paclitaxel (tax); or a combination of both (CCT + tax). Cells were treated for 72 h, washed, trypsinized and counted. The percent of viable cells relative to control was plotted for each drug or combination. **B**. Same as **A** with < IC_{50} concentration the putative SP1-binding inhibitor, mithramycin (mith). Error bars represent standard deviation of triplicates from three independent experiments. * indicates $P < 0.05$, ** indicates $P < 0.01$. 
Figure 3 Analysis of drug combinations on growth of breast cancer cells grown in 3D cultures. A. Cells were seeded on Matrigel in eight-well chamber slides as described in Materials and Methods. 3D cultures formed after two days and were treated every two to three days with single agents, vehicle (control), 1 nM paclitaxel (tax), 500 nM LY2109761 (LY), 10 μM CCT007093 (CCT), 25 nM mithramycin (mith) (upper panels) or a combination of drugs (lower panels). After 10 to 14 days, mammospheres were visualized using phase-contrast microscopy. Bar scale, 50 μm. B. To count cell numbers, the Matrigel was dissolved, mammospheres were collected, trypsinized and single cells were counted by trypan blue exclusion assay using a hemocytometer. The percent cell number relative to control was plotted for each drug or combination for the two cell lines. Error bars represent standard deviation from replicates from three independent experiments.
Figure 4 Drug combinations to enhance cell death of TNBC cell lines. A. Twenty-two triple-negative cell lines were each seeded in 96-well plates. The next day cells were treated with vehicle control or paclitaxel (0.3 to 30 nM). IC₅₀ values for each cell line were generated based on the median-effect plot from three independent experiments. IC₅₀ values represent the inhibitor concentration required for a 50% reduction in cell viability relative to vehicle-treated controls. Error bars represent standard deviation of four replicates from three independent experiments.

B. Cell lines were seeded in 96-well plates and treated with single agents (IC₅₀ values) or a combination of drugs (CCT007093 + paclitaxel or mithramycin + paclitaxel) of the IC₅₀ concentrations of each drug (1:1 ratio) serial-diluted (IC₅₀-IC₂₅-IC₁₂.₅). Combination index (CI) values were calculated using the Chou-Talalay method with CalcuSyn software (Biosoft). CI values significantly > 1 are antagonistic, not significantly different than 1 are additive, and values < 1 are synergistic. Error bars represent standard deviation of quadruplicates from three independent experiments.
assayed the two lines used in our RNAi screening, MDA-MB-231 and MDA-MB-468 for comparison (Figure 4A). A four-day cell viability assay after combination treatments was used to assess drug synergy, defined as the combination of two agents that have a greater therapeutic effect than would be expected by the addition of individual effects of each drug. The well-established Chou and Talalay method was used to determine drug synergy, as described in Materials and Methods [53]. Combination index (CI) values were derived from the median-effect plots of single agents alone or in combination and statistical tests were used to determine whether the CI values at multiple dose-effect levels (IC50, IC25, IC12.5) were statistically significantly different from 1 (P < 0.05). CI values significantly < 1 indicate synergy, not significantly different from 1 indicates additive, and a CI value significantly > 1 indicates antagonism.

CCT007093 was synergistic with paclitaxel (average CI value significantly < 1, P < 0.05) in two paclitaxel-sensitive cell lines, MDA-MB-468 and MDA-MB-231, average CI value of 0.56 and 0.38, respectively, and in two of the four paclitaxel-resistant cell lines CAL120 (average CI = 0.89) and HDQP1 (average CI = 0.65) (Figure 4B). CCT007093 was additive with paclitaxel in the two other paclitaxel-resistant cell lines SW527 and MT3 (average CI values not significantly different than 1 (P > 0.05)). Mitramycin was synergistic with paclitaxel in the two paclitaxel-sensitive lines MDA-MB-468 and MDA-MB-231, average CI value of 0.66 and 0.54, respectively, and the paclitaxel-resistant cell line HDQP1 average CI value 0.87. However, mitramycin and paclitaxel were antagonistic, average CI values significantly > 1, in reducing cell viability at high effective drug doses (IC50 and IC25) in the paclitaxel-resistant lines CAL120, SW527 and MT3 (Figure 4B). Collectively these data indicate that novel drug combinations with paclitaxel can effectively reduce cell viability of select paclitaxel-sensitive and importantly, paclitaxel-resistant TNBC cell lines.

Discussion

Our RNAi screen represents a directed approach to identifying breast cancer relevant, druggable targets to enhance drug sensitivity. The screen was validated by our finding that several of the positive hits are genes that are known targets of paclitaxel sensitivity and have been clinically targeted in combination with taxanes [54,58,61,68,69]. We identified additional novel gene targets and respective agents that were not previously identified by drug sensitivity RNAi screens or whose inhibitors were not previously combined with paclitaxel.

We found PPM1D as a target for paclitaxel sensitivity in our RNAi screens and in follow-up studies observed synergistic inhibition of tumor cell growth with use of the PPM1D inhibitor CCT007093 in high PPM1D, wild-type p53 expressing MCF-7 cells. The oncogenic activity of PPM1D expression is attributed to its phosphatase activity and ability to deregulate tumor suppressor genes such as p53, Chk1, and p38 [71]. PPM1D contributes to the development of human cancers by suppressing p53 activation and thus has been an attractive therapeutic target in tumors that overexpress PPM1D and those with wild-type functional p53 activity [73]. Indeed, others have found that suppression of PPM1D expression by RNAi inhibitors proliferation and induces apoptosis in breast cancer cell lines with wild-type p53 (BT-474) and those with PPM1D amplification (MCF-7 and ZR-75-1) [86]. However, the effect of inhibition of PPM1D on tumor cell growth and drug sensitivity is not limited to tumor cells that harbor these amplifications as we observed synergistic or additive activity of CCT007093 with paclitaxel in TNBC cell lines (MDA-MB-231 and MDA-MB-468, mutant p53 cell lines) including some paclitaxel-resistant lines. Likewise, Belova et al. identified chemical compounds that inhibit PPM1D activity and showed that these compounds could significantly inhibit tumor cell growth in MCF-7 cells and those with low PPM1D, mutant p53 expression MDA-MB-231 [74]. Interestingly, PPM1D inhibitors in both of these cell lines were able to potentiate the effects of doxorubicin but failed to enhance activity in other cell lines (MDA-MB-361) [74].

We found that mitramycin, an inhibitor of SP1 binding, could synergize with paclitaxel in some TNBC (basal-like) cell lines, MDA-MB-231, MDA-MB-468, and HDQP1. SP1 is a zinc finger transcription factor important in the regulation of genes involved in cell survival, growth and differentiation, and tumor development and progression [77]. SP1 cooperates with other prominent transcription factors including oncogenes such as MYC, which may contribute to tumor cell proliferation and growth [87,88]. MYC has recently been shown to have elevated activity and gene signatures present in basal-like TNBCs [89,90]. Thus, inhibiting SP1 binding with mitramycin may block oncogenic transcriptional activity and cooperate with anti-mitotic agents such as paclitaxel to inhibit tumor cell growth. In addition, SP1 is a potent transactivator of IGF-IR and EGFR, two prominent genes overexpressed in breast cancer cells (for example, MDA-MB-468) and both of which were identified as hits in our screen [91,92].

Despite extensive preclinical studies aimed at therapeutically targeting the TGFβ signaling pathway, there is a lack of reports in which TGFβ inhibitors are combined with paclitaxel. We found that the TGFβR inhibitor LY2109761 is synergistic with paclitaxel in breast cancer cells grown in 3D cultures but not 2D cultures, indicating the importance of performing drug combination in more than one growth context. TGFβ protects mammary epithelial cells from apoptosis in the absence of serum,
which may be through activation of the PI3K/AKT cell survival pathway [80,93]. Thus, inhibition of TGFβ may sensitize cells that are grown in low-serum and/or anchorage-independent 3D conditions to apoptosis-inducing agents like paclitaxel. In support of this, others have shown that inhibition of TGFβ in human breast carcinoma cells grown in 3D cultures that secrete high levels of TGFβ increases the cytotoxic response to ionizing radiation and several chemotherapeutic drugs, namely cisplatin [94]. Likewise, inhibition of TGFβ can prevent radiation-induced acceleration of metastatic cancer progression [95,96]. On the contrary, Ahmed et al. showed that the loss of the ECM protein TGFβI is sufficient to induce specific resistance to paclitaxel and mitotic spindle abnormalities in ovarian cancer cells [97]. In ovarian and breast tumor specimens, TGFβI expression was shown to be tightly co-regulated with other genes that induce paclitaxel sensitivity, such as the adhesion glycoprotein, THBS1 [97].

The mechanism by which inhibition of TGFβ signaling cooperates with paclitaxel is not well understood. Intracellular TGFβ signaling proteins Smad2 and Smad3 bind microtubules, and upon TGFβ stimulation, these transcription factors dissociate from microtubules, are phosphorylated and relocate to the nucleus [98]. TGFβ signaling may serve as a growth promoter and/or enabling changes in tumor cell adhesion, migration, and host-tumor interactions [99]. Thus, loss of TGFβ signaling may sensitize cells to paclitaxel, an agent that can also alter adhesion and migration due to significant changes in microtubule dynamics that are required for these biological activities.

The ever-increasing volume of genomic information paired with bioinformatic and biostatistical analyses is making genotype-driven health care a reality. The tremendous amount of tumor-derived genomic information available now, and after completion of several large-scale cancer sequencing efforts, combined with biological validation of mutations to determine relevant drivers, will allow for much more facile identification of new targets for drug discovery, as well as more precise alignment of patients with a particular targeted therapy. Validation of putative drug targets through loss-of-function screening, similar to that performed herein, will likely be a frequently used approach to generate requisite pre-clinical data to investigate novel single agent and drug combinations. The exciting challenge ahead of us is to integrate the ever-expanding genomic information as quickly as possible for human benefit.

**Conclusions**

We used a genomic-based approach that included loss-of-function RNAi screening to identify druggable targets involved in paclitaxel sensitivity in breast cancer cells. We identified pharmacological agents that target hits from our screens, several which sensitized breast cancer cells to paclitaxel. A potential translation of our discoveries is new treatment options for patients with TNBC disease, those without current clinically proven targeted therapies. In summary, we provide a platform in which integrated genomic information can be rationally used to design functional screens to identify druggable targets to improve current treatments or to discover novel cancer treatment strategies.

**Additional material**

**Additional file 1 Growth conditions of breast cancer cell lines.** Cell culturing conditions for panel of triple-negative breast cancer cell lines.

**Additional file 2 Knockdown of PPM1D expression correlates with increased paclitaxel sensitivity in breast cancer cells.** A Non-silencing (NS) control, four individually designed PPM1D-1 (P1 to P4) or pooled PPM1D siRNAs (Dharmacon ON-TARGET plus) were transfected into MCF-7 and MDA-MB-468 cells. Cells were harvested 72 h after transfection, RNA purified, and the relative PPM1D mRNA expression was measured by quantitative real-time PCR. PPM1D mRNA knockdown by individual or pooled siRNAs is shown relative to non-silencing control. Error bars represent standard deviation from three independent experiments. B Following transfection of siRNAs, as indicated above, MCF-7 and MDA-MB-468 cells were seeded in 12-well plates and treated with paclitaxel (0 to 3 nM) for two days. Cells were counted and quantified at 10 days after plating. The dose response curves of the surviving fractions are plotted relative to vehicle control treated cells. Error bars represent standard deviation of triplicate wells from three independent experiments.

**Abbreviations**

2D: two-dimensional; 3D: three-dimensional; CI: combination index; DMEM: Dulbecco’s modified Eagle’s medium; ECM: extracellular matrix; ER: estrogen receptor; IC: concentration that inhibits growth compared to control; NS: non-silencing; PR: progesterone receptor; RNAi: RNA interference; shRNA: short hairpin RNA; SI: sensitivity index; siRNA: small interfering RNA; TNBC: triple-negative breast cancer.

**Competing interests**

The authors declare that they have no competing interests.

**Authors’ contributions**

JAB and JAP designed the study, analyzed data, and prepared the manuscript. CLA contributed to the interpretation of the results and discussion. CBM, BDL, and CP assisted with the RNAi screen and drug combination experiments. FY and YS provided a model and performed statistical analysis of the data generated from the RNAi screen. All authors read and approved the submitted manuscript.

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