A platform of patient-derived microtumors identifies therapeutic vulnerabilities in ovarian cancer.

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39 Abstract

40 Background:

41 In light of the frequent development of therapeutic resistance in cancer treatment, there is a strong need for personalized model systems representing patient tumor heterogeneity, while 42 43 enabling parallel drug testing and identification of appropriate treatment responses in individual 44 patients. Using ovarian cancer as a prime example of a heterogeneous tumor disease with 45 complex microenvironment and high recurrence rates, we developed a 3D preclinical ovarian 46 cancer model comprised of patient-derived microtumors (PDM) and autologous tumor-infiltrating 47 lymphocytes (TILs) for identification of treatment vulnerabilities and validation of anti-cancer drug 48 efficacy using immunohistochemistry, immune cell phenotyping, functional assays and protein 49 profiling analyses.

50 Methods:

- 51 PDM and TILs were isolated from fresh primary ovarian cancer tissue specimen using
- 52 mechanical disruption and limited enzymatic digestion and were subsequently cultured in
- 53 suspension in defined media in the absence of serum. The heterogeneous cellular composition of
- 54 isolated PDM as well as autologous TILs was analyzed by FFPE immunohistochemistry and
- 55 multi-color flow cytometry, respectively. For in-depth protein profiling of PDM we established
- 56 Reverse Phase Protein Array (RPPA) analyses of >110 total and phospho-proteins. Treatment
- 57 efficacy in response to chemotherapeutics as well as immunotherapeutic compounds was
- assessed in PDM and PDM-TIL co-cultures using a functional viability assay in microplate format.

59 Results:

- 60 The enzymatic digestion of primary ovarian cancer tissue and suspension culture in defined
- 61 serum-free media allowed fast and efficient recovery of patient-derived microtumors (PDM).
- 62 Immunohistochemical analyses demonstrated histopathological comparability of ovarian cancer
- 63 PDM with corresponding patient tumor tissue. Reverse Phase Protein Array (RPPA)-based
- 64 analyses of >110 total and phospho-proteins enabled the identification of patient-specific
- 65 sensitivities to standard, platinum-based therapy and thereby the prediction of potential

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- 67 of immune checkpoint inhibitor treatment demonstrated the potential for patient-specific
- 68 enhancement of cytotoxic TIL activity by this therapeutic approach.

69 Conclusion:

70	3D patient-derived ovarian cancer microtumors represent a preclinical, ex vivo tumor model that
71	reflects intertumoral heterogeneity and represent the cellular complexity of individual patient
72	tumors. Combining protein pathway analysis and anti-cancer drug efficacy testing of PDM
73	enables drug mode-of-action analyses and therapeutic sensitivity prediction within a clinically
74	relevant time frame after surgery. Follow-up studies in larger cohorts are currently under way to
75	further evaluate the applicability of this platform to support clinical decision-making and
76	personalizing cancer treatment.

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78 Keywords79

Preclinical tumor model, ovarian cancer, anti-cancer drug efficacy testing, cancer immunotherapy,
protein profiling, therapy resistance

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83 Main Text

85 Background

86 In the context of personalized medicine, patient-derived model systems are expected to play an 87 important role in order to identify suitable and effective therapies for the individual patient as well 88 as existing therapeutic resistances of the patient's tumor. Especially for cancer types with dismal 89 treatment success rates such as ovarian cancer (OvCa), these model systems will be valuable for 90 future cancer therapy. OvCa is among the most lethal gynecological diseases in women with 91 >185.000 deaths worldwide in 2018 (1). Late diagnosis and disease complexity characterized by 92 strong molecular and genetic heterogeneity are causative for its poor survival rates and varying 93 treatment response to first-line therapy. Substantial efforts have been made to resolve the 94 complexity of OvCa, especially for high-grade serous carcinomas (HGSC) (2-4). Despite the 95 application of genomics and transcriptomics in elucidating disease determinants, the principles of

responsiveness to therapy are still poorly understood (4). The establishment of patient-derived 96 97 tumor organoids (PDO) allowed addressing a number of these challenges for example by in-98 depth genetic and phenotypic tumor characterization and analysis of intra-tumoral heterogeneity 99 in PDOs side-by-side with corresponding tumor tissue (5-8). Even though recent studies have 100 described the combination of PDO cultures with components of the tumor microenvironment 101 including fibroblasts, endothelial cells and immune cells (9), PDOs do not fully reflect the original 102 composition of primary tumor tissue in terms of extracellular matrix, tumor-associated fibroblasts, 103 tumor-infiltrating lymphocytes (TILs), macrophages (TAMs), and tumor endothelial cells. Another 104 challenge of current PDO models in terms of applicability for individualized drug response testing 105 relates to the required establishment time of 1-3 months with a corresponding impact on the 106 timeframe to obtain drug testing results (10). Using OvCa as a prime model of a heterogeneous 107 tumor disease, we introduce a three-dimensional (3D) preclinical ex vivo model composed of 108 patient-derived microtumors (PDM) as well as autologous tumor-infiltrating lymphocytes (TILs) 109 extracted from primary OvCa tissue specimen in a clinical relevant time-frame. Importantly, PDM 110 recapitulate a 3D histo-architecture with retained cell-cell contacts and native intra-tumoral 111 heterogeneity featuring the corresponding primary tumor microenvironment (incl. extracellular 112 matrix proteins, stromal fibroblasts and immune cells). In combination with functional compound 113 efficacy testing and multiplexed TILs phenotyping, we demonstrate the correlation of individual 114 OvCa PDM responses to chemotherapeutic as well as immunotherapeutic treatment approaches 115 using OvCa PDM alone and in co-culture with autologous TILs, respectively. We apply Reverse 116 Phase Protein Array (RPPA) analysis to map protein-signaling pathways of PDM and to measure 117 on- and off-target drug effects in compound treated PDM. Albeit based on a small patient cohort 118 the available clinical follow-up data suggests a correlation of obtained treatment responses in 119 OvCa PDM models and corresponding patients indicating prolonged metastasis-free survival of 120 identified carboplatin-responders as compared to non-responders.

121 Based on the data presented here, we envision that our preclinical assay system combining

122 PDM, autologous TILs and protein signaling pathway profiling could aid clinical decision-making

- in the future and assist in the pre-selection of a personalized clinical treatment strategy for OvCa.
- 124

125 Materials and Methods

126127 Human specimens

128 Ovarian tumor samples were obtained from nineteen patients diagnosed with ovarian cancer

129 undergoing surgery at the Center for Women's Health, University Hospital Tuebingen. Written

130 informed consent was obtained from all participants. The tumors were classified according to

131 International Federation of Gynecology and Obstetrics (FIGO) grading system. Tumor samples

- 132 were delivered on the day of operation. The research project was approved by the ethics
- 133 committee (IRB#275/2017BO2 and IRB#788/2018BO2).

134 Isolation and cultivation of patient-derived microtumors and tumor-infiltrating135 lymphocytes

The procedure was adapted from Kondo et al. (2011) (11) and modified as follows. Tumor
specimens were washed in HBSS (Gibco), minced with forceps, and digested with
LiberaseTM DH (12) for 2h at 37°C. Digested tissue was centrifuged (300g, 5 min), washed
with HBSS and filtered through a stainless 500 µm steel mesh (VWR). The flow-through was

- again filtered through a 40 µm cell strainer (Corning). The filtrate containing the TIL fraction was
- resuspended in Advanced RPMI 1640 (Gibco) supplemented with 2 mM Glutamine (Gibco), 1%
- 142 MEM Vitamins (Gibco), 5% human serum (Sigma-Aldrich) and 100 µg/ml primocin (Invivogen).
- 143 IL-2 (100 U/ml), IL-7 (10 U/ml) and IL-15 (23.8 U/ml) (Peprotech) were freshly added to culture

144 media. For expansion, CD3/CD28 dynabeads were added (Milteny Biotech). PDM, held back by

145 cell strainer, were washed in HBSS and cultured in suspension in StemPro® hESC SFM (Gibco)

- 146 supplemented with 8 ng/ml FGF-basic (Gibco), 0.1 mM β-mercaptoethanol (Gibco), 1.8% BSA
- 147 (Gibco) and 100 µg/ml primocin (Invivogen) within cell-repellent culture dish (60x15 mm)

148 (Corning).

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150 RPPA and protein data analysis

151 Detailed methods of sample preparation and RPPA processing are provided in SI Materials. 152 RPPA protein analysis and protein data processing was applied as reported before (13-15). 153 Detailed methods of sample preparation and RPPA processing are provided in SI Materials. From 154 the arrays, PDM sample signals were extracted as protein-normalized, background-corrected 155 mean fluorescence intensity (NFI), as measured from two technical sample replicates. NFI 156 signals, median-centered for each protein over all measured samples (including OvCa PDM and 157 BC PDM samples) and log₂ transformed, reflect a measure for relative protein abundance. Small 158 NFI protein signals at around blank assay level (0.02 NFI) were as a limiting quality criterion 159 excluded from further analysis; otherwise all NFI signals were used for further protein data 160 analysis. Protein heat maps were generated and cluster analysis (HCL) performed using the 161 freely available MultiExperiment Viewer (MeV) software. For the comparison of protein profiles of 162 treatment responders and non-responders (defined by functional compound testing), only 163 proteins with a >20% difference between the means were used for analysis. On- and off-target 164 pathway effects were evaluated from one biological and two technical replicate samples per 165 model at three different treatment times (0.5, 4 and 72 h). Treated sample to respective DMSO 166 vehicle control NFI ratios (TR) were calculated for each treatment condition and log2-transformed. 167 A treatment-specific threshold of protein change (carboplatin: minimum 50% difference) was set. 168 Only proteins showing treatment effects above the threshold were shown.

169 Efficacy of compounds validated in PDM cultures

Efficacy of compounds was validated by applying the real-time CellTox[™] Green Cytotoxicity
assay (Promega). Assays were performed according to manufacturer's protocol. PDM were
cultured a maximum of 1-2 weeks in PDM culture medium prior testing. Per treatment three to
eight replicates were performed using n = 15 PDM per replicate in a total volume of 150 µl
phenol-red free PDM culture medium. Cell death was measured as relative fluorescent unit (RFU)
(485–500 nm Ex / 520–530 nm Em), relative to the number of dead, permeable cells after 24h,
48h and 72h with the Envision Multilabel Plate Reader 2102 and Tecan Spark Multimode Plate

177 Reader. RFU values were normalized to DMSO control according to used drug solvent.

Treatment effects were measured as fold change (FC) compared to control. Differences between
treated PDM and untreated PDM were calculated as fold change values separately for each time
point. Statistical significance was evaluated by two-way ANOVA multiple comparison test.
Outliers were identified with the Iglewicz and Hoaglin's robust test for multiple outliers applying a

182 recommended Z-score of \geq 3.5 (16).

183 FACS analysis

To characterize lymphocyte populations within autologous TIL, cells were harvested (up to 1x10⁶ 184 185 cells/staining depending on available number of cells), washed 2x with PBS (200 rpm, 5 min at 186 4°C), resuspended in staining buffer (PBS plus 10% FBS) and plated in a 96-well V-bottom plate 187 (100 µl/well) (Corning). To verify >90% cell viability, cells were counted with a Nucleocounter 188 (Chemotec) before plating. For each panel staining, an unstained control and if necessary a FMO 189 control were prepared. For extracellular staining, cells were incubated with antibodies (see SI 190 Materials) for 30 min at 4°C in the dark. For subsequent intracellular staining, cells were washed 191 2-3 times (200 rpm, 5 min at 4°C) in eBioscience™ Permeabilization buffer (250 µl/well) 192 (Invitrogen) and resuspended in eBioscience ™ Fixation/Permeabilization solution (Invitrogen) for 193 20 min at 4°C. After 2-3 washing steps (200 rpm, 5 min at 4°C), cells were incubated with 194 antibodies (30 min, 4°C in dark) (see SI Materials). After the staining process, cells were washed 195 2-3 times and analyzed with a BD FACS Melody machine (BD Biosciences).

196 Co-culture of PDM and autologous TILs

To measure if the expanded, autologous TILs are able to kill corresponding PDM, we performed
endpoint killing assays in a 96-well format with an image-based analysis using Imaris 8.0
software. First, PDM were pretreated with IFNγ (200 ng/ml) for 24 h to stimulate antigen
presentation. In parallel, 96-well plates were coated with 5 g/ml of anti-CD28 antibody (Biolegend)
o/n at 4°C to provide a co-stimulatory signal during co-culture. On the next day, coated plates
were washed 3x with PBS. PDM were washed in HBSS, centrifuged and resuspended in co-

203 culture assay media consisting of RPMI 1640 phenol red free (GIBCO) supplemented with 2mM 204 Glutamin (Gibco), 5% human serum (Sigma-Aldrich), 1x MEM Vitamins (Gibco) and 100 µg/ml 205 Primocin (Invivogen). Prior of assembling the co-culture, TILs were labeled with CellTracker™ 206 Deep Red Dye (Thermo Fisher Scientific) to differentiate between PDM and TILs. Labeled TILs 207 were then co-cultured with PDM and in the presence of selected checkpoint immune inhibitors 208 (CPIs: Pembrolizumab, Atezolizumab, Ipilimumab; Selleck Chemicals GmbH) or control anti-IgG4 209 antibody with an E:T ratio of 4:1. Thereby we counted 200 cells per single PDM. Per condition, 210 we prepared triplicates each with 15 PDM and 12 000 TILs per well. After 92 hours, cells were 211 incubated with a staining solution consisting of live cell stain Calcein-AM (Thermo Fisher 212 Scientific) and Sytox™ Orange dead cell stain (Thermo Fisher Scientific). After 1 hour, Z-stacks 213 of n = 3 PDM per well were imaged using a spinning disk microscope (ZEISS CellObserver Z1). 214 Only viable PDM were positively stained by Calcein-AM, while all dead cells were stained by 215 Sytox™ Orange. TILs were filtered by CellTracker™ Deep Red signal. Using the Imaris 8.0 216 software, we applied three masks, one for dead cells, one for dead TILs and one for live PDMs. 217 For each mask, the total sum of all fluorescent intensities (FI) was calculated and the following 218 ratio determined:

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% ratio dead vs.viable PDM [FI] = $\frac{\text{total dead [FI]} - \text{dead TIL [FI]}}{\text{viable PDM [FI]}}$

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221 Statistical analysis

Statistical analysis was performed using GraphPad Prism. For Boxplot data, whiskers represent
quartiles with minimum and maximum values and the median. Datasets with no clear normal
distribution were analyzed with unpaired, two-tailed Mann-Whitney-U-test, otherwise as indicated.
Correlation data were evaluated by Spearmans rank correlation. For all analyses, *P* values < 0.05
were considered statistically significant. Recommended post-hoc tests were applied for multiple
comparisons.

229 Results

Isolation of patient-derived microtumors with high viability from primary OvCa tissue specimen by limited enzymatic digestion

- 232 Residual fresh tumor tissue samples were collected from n = 19 OvCa patients undergoing
- 233 primary tumor debulking surgery. Available, anonymized clinico-pathological characteristics
- 234 including International Federation of Gynecology and Obstetrics (FIGO) staging and pathological
- 235 TNM-classification of respective individuals are summarized in Table 1.

FIGO Sample Age at Histopathological Cellular origin v Isolated PDM Grade т Pn R Ν м L OvCa # surgery classification stage pT3c pN1a (2/14) HGSC HG V1 Pn1 #1 53 Ш Мx L1 Rx epithelial yes (liver/splenic capsule) pT2b #3 88 HG IIВ L0 HGSC epithelial Nx Мx V0 Pn0 Rx yes (peritoneum douglas) #4 54 HGSC HG/G3 IIIC ypN0 L0 V0 epithelial урТ3с Мx Pn0 Rx yes pN1a (2/18) #5 59 HGSC epithelial HG/G3 IIIC pT3c Мx L0 V0 Pn0 Rx no #7 67 HGSC epithelial/peritonea HG IVa Тх Nx Мx Lx Vx Rx Pnx yes #8 44 MC epithelial LG la T1a Nx Мx Lx Vx Pnx Rx yes pT3c eal me pN1a (2/6) Pnx #11 84 HGSC HG/G3 IIIC Мx Vx R1 epithelial Lx no No indication #12 81 FT Тх -chord/stromal no alignancy #13 71 LGSC epithelial LG/G2 IIIC pT3c pN0 Мx LO V0 Pn0 Rx yes 60 #16 HGSC epithelial HG/G3 IVb ypT3b vpN0 cM0 L1 V0 Pn0 R0 no 62 HGSC HG #17 epithelial IIIC pT3c pN0 Мx V0 Pn0 Rx L1 ves epithelial #18 61 HGSC HG/G3 IIIC урТ3с ypN0 cM0 LO VO R0 Pnx yes #19 60 HGSC epithelial HG/G3 IIIB pT3b pN0 cM0 L0 V0 Pn0 R0 yes pT3c (pleural effusion) 66 pM1a L0 V0 Pn0 #20 HGSC epithelial HG IVa pN1a Rx no -chord/stromal pNx #21 74 adult-type GCT IA pT1a cM0 L0 V0 Pn0 R0 yes pT3c ım met epithelial pN1a #23 71 HGSC HG/G3 IIIC рМх L1 V0 Pn0 Rx yes pN1b #24 73 HGSC HG IIIC pT3c Мx L1 V0 Pnx R0 epithelial yes (58/75) #25 54 HGSC epithelial HG IIA pT2a (tube) pN0 cM0 L0 V0 Pn0 cR0 yes #26 67 HGSC epithelial HG/G3 IIIC pT3c рNx Мx 11 V0 Pn0 ves

 Table 1. Clinical patient data from OvCa tumor specimen included into the study.

FT, fibrothecoma; GCT, granulosa cell tumor; HGSC, high grade serous carcinoma; LGSC, low grade
serous carcinoma; MC: mucinous carcinoma T: extent (size) of the tumor; N: spread to nearby lymph nodes;
M: spread to distant sites; L, lymphatic invasion; V, venous invasion; Pn, perineural invasion; R, residual
tumor; p, pathological state; c, clinical stage; y, restaged after neoadjuvant therapy; x, not assessed;

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241 2/19 patients (OvCa #4 and OvCa #18) received neoadjuvant treatment with

242 carboplatin/paclitaxel chemotherapy. The majority of included samples (n = 17) were derived from

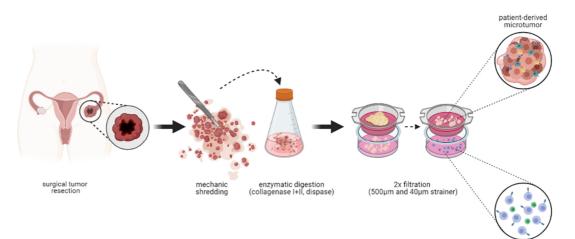
243 the most common type of OvCa, i.e. epithelial OvCa, with a majority of high-grade serous

244 carcinomas (HGSC). Two samples were classified as sex-chord-stromal ovarian carcinomas that

245 are either non-malignant or at a low stage. The PDM and TIL isolation procedure (further

246 developed from Kondo et al. 2011) (11) was performed on freshly excised tumor tissue specimen

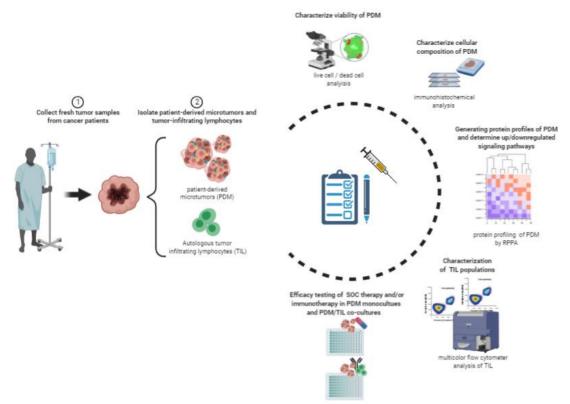
247 (Scheme 1).



Scheme 1. PDM isolation from OvCa tumor samples within 3 hours after receipt of the tumor sample. Tumor tissue derived from surgical tumor resection is kept in culture media for transportation. Immediately after receipt of the sample, the tissue is mechanically disrupted into smaller pieces and enzymatically digested for 2 hours. Afterwards the digested tissue gets filtered twice using cell strainers. Within the first filtrate tumor-infiltrating lymphocytes are obtained ready for culturing or cryopreservation. From the residue of the second strainer, PDMs are gained ready for culturing or cryopreservation. (*Created with Biorender.com*)

- 248
- 249 Isolation of PDM a was successful in >70% (14/19) of the tumor samples (Fig. 1A) with varying
- amounts of available PDM for downstream analyses such as live-dead staining,
- 251 immunohistochemical characterization, protein signaling pathway analyses and efficacy drug
- testing of standard-of-care therapy as well as immunotherapy (Scheme 2). PDMs were cultured in
- suspension in the absence of serum for a maximum of three weeks. No correlation was observed
- between successful isolation of PDM and available clinical patient data such as age, lymph node
- 255 spread, distant cancer spread, perineural invasion or FIGO stage (Table S1). PDM viability was
- assessed by parallel staining with Calcein-AM and SYTOX[™] Orange (Fig. 1B). 2D projections of
- 3D images displayed highly viable PDM with few dead cells. Dead PDM cells (according to
- 258 nuclear SYTOX[™] Orange staining) detached from PDM and thus observed mostly as single cells
- 259 floating in the culture media. The quantification of the viable cell volume and dead cell volume in

- 260 3D projections of four exemplary OvCa PDM models are shown in Fig. 1C. In each analyzed
- 261 model, ≤ 7% of the total PDM cell mass represented dead cells confirming robust PDM viability.



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Scheme 2. Illustration of established downstream analyses of PDM and TILs. After the receipt of the tumor sample, PDM and TILs are immediately isolated and cultured. The viability of PDM is checked by a live/dead cell staining. Afterwards, the total number of available PDM is assessed. PDM are paraffinembedded for immunohistochemical analyses. Anti-cancer drug efficacy is evaluated by measuring cytotoxicity in real-time requiring at least 300 PDM in total using 15 PDM per replicate (in total 4x). A minimum of 100-150 PDM are needed for further protein profiling analyses by RPPA. TILs are cultured and expanded for flow cytometry analyses as well as possible co-culture assays with PDM to assess immunotherapy sensitivity. Dependent on the total amount of PDM available, different downstream analyses are conducted and used for therapeutic sensitivity prediction. (*Created with Biorender.com*)

OvCa PDM sections resemble histopathological characteristics of the corresponding primary tumor tissue (PTT)

- 265 We next performed Hematoxylin and Eosin staining (H&E) of FFPE- and cryo-sections,
- 266 respectively, derived from OvCa PDM and corresponding primary tumor tissue sections (PTT) for
- 267 histopathological comparison. Professional assessment of PDM by a certified pathologist,
- 268 confirmed typical, distinct histopathological characteristics of OvCa in respective PDM (Fig. 2 and
- 269 Fig. S1). HGSC derived PDM reflected architectural patterns such as papillary growth, irregular
- 270 branching, cystic and glandular structures (Fig. 2 OvCa #17-23; Fig. S1, OvCa #24, 26) as

271 compared to the corresponding PTT specimen. Pleomorphic nuclei/cells, high nucleus:cytoplasm 272 ratio as wells as hyperchromasia were similar in PDM and corresponding PTT sections reflecting 273 the high-grade of analyzed HGSC tumors. These tumor features were not detected within OvCa 274 PDM #8 (Fig. S1), which originated from low-grade mucosal OvCa known for slow tumor growth. 275 Instead, OvCa #8 PDM displayed a unicellular epithelium and mostly stromal remains. In 276 summary, histopathological analyses of PDM confirmed structural and cellular similarities to the 277 corresponding primary tumor specimen resembling typical histological features of ovarian 278 carcinomas. 279 280 PDMs reflect intra- and intertumoral heterogeneity of the original tumor 281 For further characterization and study of the similarity of OvCa PDM and corresponding PTT, the 282 expression of histotype specific markers, different tumor cell-, extracellular matrix- and immune 283 cell-markers were assessed by immunohistochemistry (Fig. 2, Fig. S1). In the clinics. 284 immunohistochemical staining of p53 and Wilms Tumor 1 (WT1) is applied for differential 285 diagnosis of HGSCs (17). HGSC phenotype of the original tumor persists in the corresponding 286 PDM as marker expression of WT1/p53 corresponded well with either low-to-moderate (OvCa 287 #17-18) or strong expression (OvCa #23). Mesothelin (MSLN) and CA125 (MUC16) were used as 288 additional OvCa markers. Mesothelin, known to be over-expressed on the cell surface in OvCa 289 (18-20), was expressed in OvCa PDM mostly resembling expression of this marker in the 290 corresponding PTT. Besides MSLN, CA125 expression has previously been described as an 291 immunohistochemical marker to confirm ovarian origin of the tumor (21). As shown before (22, 292 23), expression of CA125 in OvCa sections can vary within one type and between the different 293 OvCa tumor types. Accordingly, PTT sections derived from non-HGSC displayed no CA125 294 expression (OvCa #8) in contrast to HGSC-derived tumor sections (OvCa #17-18, #23-26). In 295 PDM models OvCa#8, #17, #18 and #24 staining patterns of MSLN and CA125 were similar to 296 corresponding PTT sections and were often restricted to cells at the tumor margin for CA125 (e.g. 297 OvCa #18, #23). As the tumor microenvironment is known to play a major role in tumor 298 progression and metastasis (24-26), we analyzed the presence of extracellular matrix (ECM) and 12 299 stromal components in OvCa PDM and corresponding PTT. Sections were stained for FAPa 300 (Fibroblast associated protein alpha), a marker of activated fibroblasts, also known as cancer-301 associated fibroblasts (CAFs). FAPa expression in tumor stroma is observed in 90% of human 302 cancers of epithelial origin and has been described to induce tumor progression and 303 chemoresistance (27). FAPa staining in OvCa PDM mostly resembled that of the corresponding 304 PTT in all samples except for OvCa #24-26 and differed from low to high expression. Collagen I 305 expression, known to promote invasiveness and tumor progression in epithelial OvCa (28), was 306 also prominent within OvCa PDM. Stainings correlated with corresponding PTT that showed an 307 overall strong Collagen I expression except for OvCa #25. We also observed a strong correlation 308 of Hyaluronan Binding Protein 1 (C1QBP) expression in PDM and corresponding PTT, which 309 interacts with the major ECM component hyaluronan (29). In summary, all studied stromal 310 components were found to be highly abundant within OvCA PDM and corresponding PTT. To 311 further examine tumor microenvironmental (TME) components of PDM, we studied the infiltration 312 with tumor-associated macrophages (TAMs) via CD163 expression together with the expression 313 of the inhibitory checkpoint receptor ligand PD-L1. Analysis rarely detected M2-like TAMs 314 (CD163⁺) within PTT and PDM sections and if so, mostly in stromal tissue parts. While 315 macrophages were highly frequent in OvCa #24 PTT, they were not detected in the 316 corresponding PDM (Fig. S1). In contrast, for OvCa #17, CD163⁺ TAMs were detected in both 317 PDM and PTT sections (Fig. 2). Immune checkpoint receptor ligands are known to be expressed 318 on tumor and/or immune cells of the tumor microenvironment. Here, PD-L1 expression was mostly absent in OvCa PTT and corresponding PDM sections. In conclusion, PDM largely 319 320 resembled features of the original tumor by direct comparison with corresponding PTT including 321 presence of complex ECM and TME architecture and expression of markers specific for OvCa. In 322 comparison with corresponding PTT sections, pure stromal areas were mostly absent from 323 stained PDM sections, which might explain differences observed between PDM and 324 corresponding PTT with regard to immune cell infiltration and degree of expression of stromal 325 components.

326 Protein signaling pathway profiling of OvCa PDM by RPPA

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328 After initial immunohistochemical characterization of the 3D OvCa PDM that confirmed the 329 presence of TME components in PDM similar to corresponding PTT, we performed an in-depth 330 examination of the heterogeneity and molecular composition of different OvCa PDM models by 331 generating signaling pathway protein profiles using RPPA. Protein abundances of 116 different 332 proteins (including total and post-translationally modified forms) were measured in OvCa PDM 333 samples each with a sample size of n = 100-150 per individual PDM (Fig. 3A). One further PDM 334 sample derived from human BC (breast cancer) was included to scale up the protein sample data 335 and for comparison as both cancer types are known to share molecular and microenvironmental 336 similarities (26, 30). Obtained protein-normalized, background-corrected mean fluorescence 337 intensity (NFI) signals were median-centered to all samples (n = 8) and log_2 transformed. Protein 338 profiles of PDM samples covered signaling pathways such as for cell cycle, DNA damage 339 response, apoptosis, chromatin regulation, MAPK/RTK, PI3K/AKT with mTOR, Wnt and NFkB, as 340 well as OvCa tumor/stem cell markers. By hierarchical clustering (HCL), PDM samples were 341 grouped according to their similarities in relative protein signal intensity (Fig. 3A). Data analysis 342 revealed three clusters: 1) OvCa #21 (OvCa granulosa cell tumor) and #23 (HGSC), with the 343 most distinct protein profiles as compared to the other PDM analyzed; 2) OvCa #19 (HGSC) and 344 the BC PDM shared more similarities than OvCa #19 with the other OvCa PDM models; 3). The 345 remaining PDM samples resembled the third cluster with the most similar protein expression 346 profiles containing exclusively HGSC models. Long distances of the sample dendrogram further 347 underlines the proteomic heterogeneity of similar histopathological OvCa tumor types. 348 To compare protein abundances within different signaling pathways as well as of tumor/stem cell 349 markers, proteins with impact on pathway activity were sorted according to their pathway 350 affiliation (Fig. 3B, Table S2). Significant differences between PDM models were observed for the 351 cell cycle pathway and the MAPK/RTK pathway. Highest cell cycle activity was found in OvCa 352 #17 and #24 with almost 50% higher median NFI signals compared to OvCa #21 with the lowest 353 median signals (median NFI = -0.33 log₂) resembling a different histopathological tumor type

compared to the other PDM models analyzed. MAPK/RTK pathway signaling was increased in

355 OvCa #21 (median NFI = 0.38 log₂), #23 (median NFI= 0.32 log₂), #24 (median NFI = 0.31 log₂) 356 and #17 (median NFI = $0.30 \log_2$). The BC PDM model was characterized by decreased median 357 NFI signals of MAPK/RTK proteins (median NFI = -0.47 log₂). Proteins related to PI3K/AKT 358 pathway and of associated pathways were more abundant in OvCa #17 and #24. mTOR pathway 359 levels were elevated in OvCa #24 (median NFI = $0.54 \log_2$) in other OvCa PDM this pathway 360 showed comparable activity. Median NFI signals from apoptosis-related proteins were 361 significantly different between OvCa #25 (median NFI = 0.75 log₂) and BC PDM (median NFI= 362 1.41 log₂). OvCa tumor/stem cell marker protein abundance was significantly upregulated in both 363 OvCa #17 and #23 compared to BC PDM. Thus, RPPA protein profiling analysis demonstrated 364 the heterogeneous activity of several signaling pathways within different OvCa PDM. Apoptosis-365 related proteins and OvCa tumor/stem cell marker proteins indicated the strongest differences 366 between OvCa PDM models and the BC PDM model.

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Heterogeneous treatment responses towards chemotherapy assessed by functional compound testing in OvCa PDM 370

371 Next, we used OvCa PDM as a platform for functional compound efficacy testing using standard-372 of-care chemotherapy treatment. Further, we sought to relate treatment responses in PDM to the 373 generated protein profiling datasets. PDM were treated with carboplatin at up to three different 374 concentrations (75, 100 and 125 μ M), chosen according to the previously reported C_{max} 375 concentration of 135µM (31). Treatment efficacy in OvCa PDM – as measured by cytotoxicity -376 was heterogeneous among individual PDM models. Carboplatin induced the most significant 377 cytotoxic effects at the lowest dose (75 µM) and longest treatment duration t = 72 h in OvCa #17 378 and #24 (Fig. 3C). On the molecular level, RPPA protein profiling revealed significantly increased 379 cell cycle activity in both models (Fig. 3B), which might be associated with the comparably strong 380 carboplatin response observed in OvCa PDM #17 and #24. OvCa #23 displayed reduced 381 carboplatin sensitivity, with significant response observed after treatment at the highest dose of 382 125µM. Accordingly, OvCa #23 showed intermediate cell cycle activity in protein profiling

analyses (Fig. 3B). In conclusion, functional compound testing further confirmed the molecular

384 heterogeneity of studied OvCa PDM models identified by protein profiling, which manifests in

385 individual treatment responses to carboplatin.

386

387 Correlation of carboplatin treatment response and activation state of protein signaling 388 pathways 389

- 390 To relate the analyzed protein signaling pathways of untreated OvCa PDM to observed treatment
- 391 responses, protein NFI signals of PDM were grouped into responder and non-responder profiles
- 392 according to significant treatment effects from functional compound testing (Fig. 3C). Mean
- 393 protein signals (NFI) with >20% difference between responder and non-responder were plotted
- 394 as heat map and significant differences of pathway signaling were analyzed. Further, we
- 395 examined the on- and off-target pathway effects within different OvCa PDM models by RPPA to
- 396 assess drug mode-of-action. For this aim, OvCa PDM were treated at one compound
- 397 concentration and compared to vehicle (DMSO) control. Treatment-to-control signal ratios (TR)
- 398 were determined from protein NFI signals of treated PDM samples and DMSO vehicle controls at
- three different time points for each treatment: immediate (30 min), early (4 h) and late (72 h). This
- 400 enabled the exploration of fast and late treatment response based on changes of protein
- 401 abundances within a given time frame.
- 402
- 403 404 405

Carboplatin treatment sensitivity of OvCa PDM correlates with high protein abundance of G2-M cell cycle proteins

HCL clustering of PDM protein NFI signals led to five clusters that distinguish carboplatin 406 407 sensitive and resistant PDM models (Fig. 4A). To analyze significant differences related to 408 activation or inactivation of signal transduction pathways, proteins from the HCL clustering were 409 sorted according to their pathway affiliation and according to upregulation or downregulation in 410 responder PDM models. Carboplatin-responder PDM models showed significantly increased cell 411 cycle activity (P < 0.001; Fig. 4B) with upregulated protein abundance observed for Aurora A 412 kinase (mean NFI = 0.74 log₂), CDK2 (mean NFI = 0.8 log₂), Cyclin B1 (mean NFI = 0.84 log₂), 413 PCNA (mean NFI = 0.84 log₂), and acetylated Tubulin (mean NFI = 0.1 log₂) (Fig. S2A), which

414 are mostly related to "mitosis" (32, 33). Aurora A (Spearman's r = 0.912, P = 0.022), Cyclin B1 415 (Spearman's r = 0.971, P = 0.011) and PCNA (Spearman's r = 0.912, P = 0.022) significantly 416 correlated with carboplatin treatment sensitivity (Table S3), which was graded according to 417 recorded significance levels from "0-3" ("0": *P* > 0.05; "1": *P* < 0.05; "2": *P* < 0.01; "3": *P* < 0.001; 418 Fig. 3C). At the same time, Carboplatin non-responder PDM models showed higher abundance of 419 CDK1 (mean NFI = 0.38 log₂), phospho-CDK2 (mean NFI = 0.77 log₂) and phospho-CDK4 (mean 420 NFI = $0.37 \log_2$) (Fig. S2B), which are more related to the G0/G1 cell cycle phase. In addition, the 421 apoptosis/DNA damage response pathway was significantly upregulated in Carboplatin-422 responder compared to non-responder PDM models (P = 0.021; Fig. 4B), especially with high 423 abundance of cleaved Caspase-8 and cleaved PARP (Fig. S2A). Additional significant differences 424 between Carboplatin responder and non-responder OvCa PDM were detected within the RTK 425 and the PI3K/AKT/NF κ B signaling pathways (P < 0.001; Fig. 4B). These pathways were 426 downregulated in the Carboplatin non-responder group. Higher EMT/tumor/CSC marker 427 abundance was significantly associated with the Carboplatin responder group (Fig. 4B) including 428 protein markers Mesothelin, Nanog, STAT1, and E-Cadherin (Fig. S2A). In contrast, there were 429 few proteins found, which were down-regulated in the carboplatin responder group. Collectively, 430 this panel of down-regulated proteins differed significantly compared to the Carboplatin non-431 responder group (Fig. 4B). It contained early cell cycle markers e.g. Aurora A and Cyclin B1, the 432 mTOR pathway effector phospho-S6RP, PDGFR and SNAI1. We further assessed metastasis-433 free-survival (MFS) between the described carboplatin responder (OvCa #17, #23-25) and non-434 responder (OvCa #19, #26) PDM models (Table S4). Metastasis-free-survival (MFS) analyses of 435 available clinical follow-up patient data revealed prolonged median MFS of 16.2 month in 436 Carboplatin responder vs. versus 9.2 months in Carboplatin non-responder models. 437 In summary, the activation state of different signaling pathways comprised of proteins with >20% 438 difference in abundance, allowed to significantly distinguish carboplatin responder from non-439 responder OvCa PDM models. Importantly, these protein signaling response profiles were well in

440 line with results from functional compound efficacy testing assays using those OvCa PDM

441 models.

442 Carboplatin treatment is associated with early induction of stress-response and 443 late apoptosis 444

445 Next, we sought to investigate the carboplatin drug mode-of-action within OvCa PDM. Therefore, 446 the carboplatin-responding OvCa PDM #24 was treated with carboplatin at a concentration of 447 75 µM, which had significantly induced PDM cytotoxicity in this model (see Fig. 3C). Protein NFI 448 signals were measured at three different time points and normalized to vehicle control. Proteins 449 revealing >50% difference in TR signals (Fig. S3) were selected to focus on the strongest 450 changes in abundance. Cell cycle progression proteins (phospho-CDK2, CDK1) and phospho-451 Histone H3 (Ser10), affecting chromatin condensation during cell division, were downregulated 452 quickly within 30 minutes (Fig. 4C). After 4 h of treatment, TR signals of phospho-Aurora A/B/C 453 protein and Histone H3 was strongly increased (Fig. S3). Longer incubation with carboplatin (72 454 h) resulted in strong downregulation of these proteins (Fig. 4C). Diminished abundance of cell 455 cycle proteins after 72 h of carboplatin treatment differed significantly from vehicle control (P <456 0.001) and from early treatment (4 h; P < 0.001). While levels of cell cycle related proteins 457 decreased over time, apoptotic markers such as cleaved-Caspases as well as acetylated p53 458 were elevated after 72 h (Fig. S3). Induction of apoptosis-related proteins was already observed 459 after 4 h of treatment (Fig. 4C) with increasing abundances of cleaved Caspase 7 and acetylated 460 p53 (Fig. S3). Focusing on down-stream PI3K/AKT/mTOR/Wnt pathway regulation, the 461 abundances of mTOR effector proteins (S6RP, S6RP-phospho) were quickly upregulated after 462 immediate (0.5 h) carboplatin treatment (Fig. S3), which is in line with previous reports about 463 transcriptional regulation of stress response by the mTOR pathway (34). We also observed 464 additional elevation of mTOR pathway-related proteins after 4 h of carboplatin treatment. 465 Furthermore, overactive mTOR signaling might have resulted in increased p53 activation through 466 upregulated acetylated p53 levels (Fig. S3) as described before (34). The PI3K/AKT/mTOR 467 pathway was significantly upregulated within 4 h of carboplatin treatment compared to vehicle 468 control (P = 0.021; Fig. 4C). Similar to proteins related to cell cycle, this pathway was completely 18 469 abrogated as compared to vehicle control after 72 h of treatment (P < 0.001; Fig. 4C). Changes in 470 protein abundance differed significantly during all three measured time points (0.5 h vs. 4 h: P = 471 0.003; 4 h vs. 72 h and 0.5 h vs.72 h: P < 0.001 Fig. 4C). Pronounced, significant downregulation 472 of MAPK/RTK pathway occurred after 72 h of treatment (P = 0.017; Fig. 4C). The observed 473 proteomic changes within MAPK/RTK-related proteins over time were significant (0.5 h vs. 4 h: P 474 = 0.009; 4 h vs. 72 h: P < 0.001; Fig. 4C). Thus, carboplatin treatment of OvCa #24 illustrated 475 substantial and time-dependent changes in TR signals. Short treatment with carboplatin 476 apparently triggered the induction of stress responses while longer treatment duration caused the 477 induction of apoptosis. 478 479 Characterization of tumor-infiltrating lymphocyte populations from primary OvCa tissue 480 samples 481 482 Our established procedure of tissue processing and PDM isolation enabled us to obtain single-483 cell suspensions containing tumor-infiltrating lymphocytes (TILs) from respective OvCa tumor 484 specimen. This allowed for expansion of these autologous TILs in the presence of low-dosed 485 cytokines and antigenic stimulation in order to investigate immuno-phenotypes of respective 486 patient samples. The immunogenicity of OvCa has been demonstrated in prior studies and is 487 confirmed by the frequent infiltration of ovarian tumors with TILs (35-37). As reported by Sato et 488 al. (2005), different T cell populations diversely influence tumor immunosurveillance in OvCa. 489 High intraepithelial CD8⁺/CD4⁺ T cell ratios in patients were associated with improved survival as 490 CD4⁺ T cells executed immunosuppressive functions. To determine the composition of the 491 isolated immune cell infiltrate within our sample cohort, we characterized the phenotype of 492 autologous TIL populations by multi-color flow cytometry (Fig. S4A). Within isolated and 493 expanded OvCa TIL populations from different specimen, we found that the proportion of CD4+ 494 TILs was 57.8% and significantly more abundant than CD8⁺ TILs with 33.5% ($P = 0.003^{**}$; Fig. 495 5A, Table S5). 496

497

Isolated CD8⁺ OvCa TILs are comprised of tumor-specific CD39⁺, stem-like CD39⁻PD1⁺ and terminally differentiated CD39⁺PD1⁺ populations

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502 Within the isolated CD8⁺ TIL populations, we identified different phenotypes according to 503 expression of the co-inhibitory receptors PD-1 and CTLA-4, the tumor-antigen specificity marker 504 CD39 and the activation marker CD137 (Fig. 5A). To investigate the activation status of CD8+ 505 TILs, cells were examined for co-expression of the co-stimulatory receptor CD137 (4-1BB). 506 CD137 is upregulated in activated T cells and has been suggested to be a marker for antigen-507 activated T cells (38). The mean percentage of CD8⁺ CD137⁺ TILs was 3.1% and varied between 508 0-10% (Table S5). >5% of the CD8⁺ cytotoxic T-cells (CTLs) from OvCa #1, #3 and #5 (Fig. 5B) 509 co-expressed CD137 indicating their pre-exposure to tumor antigens. Expression of co-inhibitory 510 receptors PD-1 and CTLA-4 on CD8⁺ TILs did not differ significantly among analyzed TIL 511 populations but tended to higher PD-1 expression levels (mean 6.9% vs. 3.4%; Table S5). TILs 512 from OvCa #3, #7 and #25 as well as #5, #13 and #26 were among those displaying an 513 exhausted phenotype with >10% of CD8+PD-1+ or CD8+CTLA-4+ TILs (Fig. 5B). Moreover, in 514 recent reports CD39 expression in CD8⁺ TILs was described as a marker for tumor-antigen 515 specific TILs that have undergone tumor-antigen-driven clonal expansion, exhibit resident 516 memory T cell like phenotypes and express a variety of co-stimulatory and co-inhibitory receptors 517 (39-41). Here, CD39⁺ CTLs (mean 40.5%; range 4.4-96.8%, Table S5) were significantly 518 more abundant than CD39⁻ CTLs (mean 9.5%; range 0-48.3%, Table S5) so called 'bystander 519 TILs' known to recognizing mostly viral antigens (40) (P < 0.001, Fig. 5A). The amount of CD39⁺ 520 TILs strongly correlated with the amount of CD8⁺ TILs (Spearman r = 0.88, Fig. S4B; P < 0.001, 521 Table S6) and conversely with the amount of CD4⁺ TILs (Spearman r = -0.80, Fig. S4B; P =522 0.002, Table S6). Thus, the abundance of CD4⁺ and CD8⁺ TILs appeared to significantly 523 determine the amount of CD39⁺ CTLs. In addition, CD39 expression was largely limited to CD8⁺ 524 TILs. As co-inhibitory receptors play a role in T cell exhaustion and are important targets for 525 immune checkpoint-inhibition, we analyzed PD-1 and CTLA-4 expression on the tumor-specific 526 CD39⁺ CTL population. PD1⁺CD39⁺ were more frequent than CTLA-4⁺ CD39⁺ (15.7% vs. 5.4% 527 Fig. 5A, Table S5). The total amount of CD8⁺PD1⁺ TILs thereby correlated with the amount of 20 528 CD8⁺CD39⁺PD1⁺ TILs (Spearman *r* = 0.79, Fig. S4B; *P* = 0.002, Table S6) of a PDM model.

- 529 Thus, CD39 expression was limited to tumor-antigen stimulated and exhausted TILs (e.g. OvCa
- 530 #7, #17 and #25; Fig. 5B). In contrast to 'terminally differentiated cells' (42), OvCa TILs with a
- ⁵³¹ 'stem cell-like' CD39⁻PD1⁺ phenotype were found in 7.3% of the CTLs (Table S5). This population
- showed the highest proportional variability with a maximum of 50.5% cells vs. a minimum of 0%
- as compared to other CD8⁺ TIL populations (CV 208%). The frequency of CD8⁺CD39⁺ and stem
- 534 cell-like CD8⁺CD39⁻PD1⁺ was negatively correlated (Spearman r = -0.63, Fig. S4B; P = 0.024,
- Table S6). These results confirm the feasibility of extracting and expanding TIL populations from
- 536 fresh OvCa tissue samples and identify heterogeneous, patient-specific immuno-phenotypes with
- 537 potential relevance for immuno-oncological treatment approaches.
- 538
 539 Specific TIL phenotypes isolated from OvCa tumor specimen correlate with regional lymph
 540 node metastasis
- 542 We further analyzed the correlation between specific TIL populations and clinical follow-up patient
- 543 data. OvCa patients with regional lymph node metastasis (N = 1) tended to present with
- significantly more extensive CD8⁺ TIL infiltration in their tumors than those with no lymph node
- 545 metastasis (N = 0) (P = 0.016) (Fig. 5C). Moreover, the frequency of CD8⁺ TILs appeared to
- significantly correlate with that of CD8⁺CD39⁺ TILs in OvCa (Fig. 5C). Despite a small sample
- 547 size, our data implicate a significant correlation between lymph node spread (N=1) and the

548 presence of a CD8⁺CD39⁺ population (P = 0.008).

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550 OvCa PDM killing by autologous TIL populations is enhanced by immune checkpoint 551 inhibitor treatment 552

553 To evaluate the functional, tumor cell killing capacity of autologous TILs on OvCa PDM and the

- 554 corresponding treatment efficacy of established immune checkpoint inhibitors (CPI), we subjected
- 555 co-cultures of TILs and PDM from OvCa #24 and #26 to image-based analysis of
- 556 CPI-treatment response. A total of nine PDM were imaged per treatment (3 PDM per well in
- 557 triplicates) and a dead:live PDM ratio was calculated according to the sum of measured
- 558 fluorescent intensities (FI) (Fig. 5D-F). Addition of TILs to autologous PDM induced a basal killing

559 effect in PDM in both models analyzed in the absence of CPI treatment (Fig. 5E-F). As the 560 addition of matched isotype controls showed no additional effect in both co-culture models, we 561 excluded the possibility of unspecific killing effects of CPI antibodies. TIL killing effects in OvCa 562 #24 co-cultures were observed in response to treatment with either the combination of anti-PD1 563 and anti-CTLA-4 (pembrolizumab + ipilimumab) or anti-PD-L1 and anti-CTLA-4 (atezolizumab + 564 ipilimumab) (P = 0.039) compared to isotype control treatment (Fig. 5E). Single agents induced 565 no significant increase in PDM killing. In OvCa #26 CPI treatment almost doubled PDM killing 566 (Fig. 5F). In comparison, co-cultures treated with ipilimumab (P = 0.004) or atezolizumab (P < 0.004) 567 0.001) showed significant PDM killing compared to untreated PDM. The killing effect of TILs was 568 significantly amplified by atezolizumab treatment compared to co-culture controls (PDM+TIL: P = 569 0.021; PDM+TIL+IgG4: P = 0.018; Fig. 5F), In line with this observation, respective OvCa PDM 570 models showed weakly positive PD-L1 staining (Fig. S1). Further, atezolizumab treatment 571 significantly increased the TIL killing effect towards PDM as compared to pembrolizumab (P =572 0.026). Autologous CD8 TILs from both tested OvCa PDM models were composed of high 573 amounts of tumor-specific, non-terminally differentiated CD8⁺CD39⁺ TIL populations as compared 574 to other OvCa TILs (Fig. 5B). Moreover, these CD8 TILs were prominently positive for CTLA-4, 575 which might explain the observed increase in PDM killing in response to ipilimumab (Anti-CTLA4) 576 treatment (Fig. 5B) Thus, the co-culture of autologous TILs and PDM offers the possibility to 577 extent compound efficacy testing beyond chemotherapeutic compounds to immune oncological 578 treatment approaches in a patient-specific setting.

579 **Discussion** 580

Recently we could show the establishment of PDM from human glioblastoma tissue specimen containing important components of the tumor stroma (e.g. tumor-associated macrophages), and their application for the assessment of responses towards CSF1R- and PD1-targeting antibodies as well as the small molecule inhibitor Argyrin F (43, 44). In the present study, we have now further extended this approach to a patient-derived model system composed of PDM and autologous TILs extracted from a panel of primary OvCa tissue specimen and their in-depth

587 characterization by immunohistochemistry, protein profiling, immune cell phenotyping and 588 focused compound efficacy testing. Our results show a 70% success rate for isolation of PDM 589 with robust viability and in suitable amounts for further, multi-parametric downstream analyses. 590 In-depth histopathological assessment of PDM sections confirmed the conservation of typical 591 histological features of respective OvCa types by this model system. Importantly, the complexity 592 of the ovarian cancer TME with respect to the presence of cancer-associated fibroblasts and 593 extracellular matrix components incl. collagen and hyaluronan-binding protein observed in 594 primary OvCa tissue sections was well conserved in PDM. The presence of these TME 595 components has previously been correlated with tumor stage, prognosis, and progression and 596 shown to substantially influence treatment responses (28, 45, 46). Interestingly, we could also 597 identify immune cell infiltration within a subset of OvCa PDM, reflecting the immunogenicity of 598 OvCa as previously reported (35, 36, 47).

599 While OvCa patient-derived organoids (PDO) were often studied by genomic and transcriptomic 600 sequencing (6-8), we were the first (to our knowledge) to investigate inter-tumoral heterogeneity 601 and differential drug response mechanisms by RPPA-based protein profiling in a patient-derived 602 3D OvCa preclinical cell model. Here, analyses of a panel of >110 phospho- and total proteins 603 allowed for the clustering of histologically similar OvCa PDM models, pathway activity profiling 604 and investigation of on- and off target drug effects. Obtained RPPA protein profiles confirmed the 605 heterogeneity of OvCa PDM observed via immunohistochemistry and previously reported for 606 HGSC, the most common type of OvCa. Our work identified significant differences in the activity 607 of cell cycle and MAPK/RTK pathways within analyzed OvCa PDM and enabled their distinction 608 from a breast cancer derived PDM model by differential expression of OvCa tumor and stem cell 609 markers as well as apoptosis-related proteins.

Six OvCa PDM models were applied for individualized compound efficacy testing of standard of
care platinum-based therapy. For analyzed OvCa PDM models, we observed patient-specific
heterogeneity of response towards chemotherapy. Correlation with RPPA protein profiling data
allowed the allocation of individual PDM drug responses to specifically up- or down-regulated

614 signaling pathway activities and enabled the prediction of PDM models with high probability of 615 response towards chemotherapy. In accordance with the ability of cytostatic drugs to induce 616 apoptosis especially in actively dividing cells (48), our work identified additional correlation 617 between proteins relevant for S- and G2/M- cell cycle phase progression and carboplatin 618 response. Specifically, our data implicate that elevated abundances of Aurora A, Cyclin B1 and 619 PCNA proteins may allow for identification of carboplatin treatment response. Furthermore and in 620 line with previous reports, we confirmed that decreased DNA damage repair and the ability to 621 undergo apoptosis (49) is associated with carboplatin treatment sensitivity in OvCa. This was 622 illustrated by increased levels of cleaved Caspase-7 and cleaved PARP. Our results did not 623 identify a correlation of carboplatin-resistance and markers of cancer stem cells (CSCs) (50, 51) 624 or epithelial-to-mesenchymal transition (EMT) (52, 53). Instead, we found the cancer stem cell-625 related protein Nanog as well N-Cadherin strongly upregulated in carboplatin responding PDM. 626 These differing results might arise from the fact that above-mentioned previous studies were 627 performed in adherent cell lines and not within a patient-derived 3D tumor model. Importantly, we 628 identified protein signatures of OvCa PDM allowing for the identification and prediction of PDM 629 models with high probability of response towards chemotherapy or targeted therapy. The 630 correlation of our results with clinical data indicated a significant correlation of carboplatin 631 treatment response with prolonged metastasis-free survival of respective patients. Given the 632 small sample cohort analyzed here, these results need to be interpreted with caution but warrant 633 further investigation. 634 We further assessed proteomic changes upon PDM treatment such as effects on protein

abundance, directed on- and off-target pathway effects and drug mechanism-of-action within
OvCa PDM. In a carboplatin sensitive PDM model, we observed a time-dependent decrease in
cell cycle- and an increase in apoptosis-inducing protein abundance. In parallel, we found a fast
stress response upon treatment as indicated by an activated mTOR pathway with high S6RP and
active phospho-S6RP levels (34). Overactive mTOR in combination with cell stress and the

640 inability of cells to adapt to cellular stress might be responsible for p53 elevation (54, 55) and641 driving cells into senescence or apoptosis (56, 57).

642 Apart from testing the response of OvCa PDM to conventional chemotherapy, we sought to 643 investigate the applicability of this model system for efficacy assessment of immuno-oncological 644 treatment approaches. For this aim, we applied immunophenotyping of autologous TIL 645 populations followed by their co-culture with respective PDM in the presence and absence of 646 immunotherapeutic mono- and combination treatment schedules. Immunosurveillance of cancer 647 strongly depends on the composition of tumor-infiltrated immune cells and the degree of tumor 648 tissue infiltration and is known to influence treatment efficacies. As a result, the idea of an 649 immunoscore, identifying a patient's immunophenotype, emerged (58). Our work uncovered 650 several immunophenotypes within expanded TILs from OvCa patients by multicolor flow 651 cytometry compared to previous immunohistochemistry based analysis (59). As described by 652 Sato et al. (2005) (35) and Zhang et al. (2003) (36) high numbers of intraepithelial CD8⁺ TILs are 653 associated with better prognosis in OvCa. We found that OvCa TILs were largely composed of 654 CD4⁺ rather than CD8⁺ TILs. In this regard, OvCa models with high amounts of suppressive CD4⁺ 655 TILs and low numbers of CD8⁺ TILs are suggested to have worse prognosis (60). In line with 656 previous reports (61), we identified expression of CD39 in OvCa TIL populations, a marker that 657 distinguishes between tumor-specific CTLs (CD39⁺) and bystander TILs (CD39⁻) (39, 40). 658 Interestingly, we found that CD8⁺ TIL amounts correlated with that of CD8⁺ CD39⁺ TILs, and could 659 confirm that these tumor-specific T cells constitute an exhausted, memory T cell like phenotype, 660 as CD39 expression was limited to CD8⁺PD-1⁺ TILs. Importantly, our results further demonstrated 661 that co-cultures of PDM and autologous TILs could be applied to assess treatment efficacy of 662 CPIs in a patient-specific setting. Such PDM-TIL co-culture systems could potentially be used to 663 identify OvCa patients more likely to respond to cancer immunotherapy. In the limited OvCa 664 tumor tissue cohort investigated here, OvCa tumors with regional lymph node metastasis 665 contained higher numbers of CD8⁺ and CD8⁺CD39⁺ TILs. The co-culture models tested in our

666 study for response towards CPI treatment were derived from lymph-node spreading primary 667 tumors, which might suggest that immunogenicity of OvCa increases upon metastasis. 668 Limitations of our PDM model are currently the restricted number of PDM available from digestion 669 of individual tumor tissue samples. From experience with different tumor types, an average of 670 several hundred to several thousand microtumors can be isolated from fresh tissue samples. This 671 number depends on the amount of tissue available for PDM isolation as well as tissue 672 composition (including degree of fibrosis and necrosis). PDMs are therefore presently not suitable 673 for high-throughput drug screening approaches, but for focused drug testing in late preclinical and 674 translational drug development as well as in the context of precision oncology. Furthermore, the 675 limitation of our present study with regard to sample size should be noted. In future follow-up 676 studies with larger cohorts we plan to confirm our findings. 677 678 Conclusions 679 680 Patiend-derived microtumors isolated from OvCa tumor specimen represent a novel ex vivo 681 tumor model for OvCa displaying histopathological similarities to corresponding primary patient 682 tumors and revealing intertumoral heterogeneity as evidenced by immunohistochemical and

683 protein profiling analyses. The combination of functional drug testing with analyses of protein

684 signaling pathways and drug-mode of action enabled the identification of PDM models

685 susceptible to platinum-based treatment and allowed for the prediction of individual therapeutic

686 sensitivity. Parallel isolation and culturing of autologous TILs further allowed for the

687 characterization of patient-individual immune-phenotypes as well as the assessment of

responses towards immunotherapy in PDM-TIL co-cultures. While the rapid PDM/TIL extraction

689 procedure and quick availability of resulting datasets within 3-4 weeks is in good accordance with

timelines of clinical decision making, we plan to confirm our findings in future studies with larger

691 sample cohorts.

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696 List of abbreviations

697 698	PDM:	Patient-derived microtumors
699	TIL:	Tumor-infiltrating lymphocytes
700	FFPE:	Formalin-fixed paraffin-embedded
701	RPPA:	Reverse Phase Protein Array
702	OvCa:	Ovarian cancer
703	HGSC:	High-grade serous carcinoma
704	PDO:	Patient-derived organoids
705	TAM:	Tumor-associated macrophages
706	FIGO:	International Federation of Gynecology and Obstetrics
707	IL-2/7/15:	Interleukine-2/7/15
708	FGF-basic:	Basic fibroblast growth factor
709	BSA:	Bovine serum albumin
710	DMSO:	Dimethylsulfoxid
711	BC:	Breast cancer
712	HCL:	Hierarchical Clustering
713	MeV:	MultiExperiment Viewer
714	RFU:	Relative fluorescent unit
715	FC:	Fold change
716	PTT:	Primary tumor tissue
717	H&E:	Hematoxylin and Eosin staining
718	WT1:	Wilms tumor 1
719	MSLN:	Mesothelin
720	CA125:	Cancer-antigen 125
721	ECM:	Extracellular matrix
722	FAP α:	Fibroblast activation protein α
723	CAF:	Cancer associated fibroblast

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724	C1QBP:	Hyaluronan binding protein 1
725	TME:	Tumor microenvironment
726	CPI:	Checkpoint-inhibitors
727	CSC:	Cancer-stem cells
728 729 730 731 732 733 734 735 736 737 738 739 740 741 742 743 744 745 746 747 748 749 750 751 752 753 754 755 756 757 758 759 760 761 762 763 764 765 766 767 768 769 770 771 772 773		

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Declarations

775 776 777	Ethics Approval and consent to participate
778	The study was conducted according to the guidelines of the Declaration of Helsinki and approved
779	by the local Ethics Committee. The use of human samples was approved by the local Ethics
780	Commission at the Medical Faculty of Tuebingen under the reference numbers
781	IRB#275/2017BO2 and IRB#788/2018BO2. All patients enrolled gave their informed consent to
782	participate in the study.
783 784 785 786 787	Consent for publication: Not applicable.
788 789	Availability of data and material:
790 791	All data needed to evaluate the conclusions of the paper are included in this published article and
792	its supplementary information file. Material and further data are available upon request after
793	signature of an MTA from the corresponding authors.
794 795	Competing interests:
796	AH received consulting and speaking fees from GSK, AstraZeneca and Clovis. NA, AK, BG, ALK,
797	AS, SYB, MP, KSL and CS declare no competing interests.
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802 803	Authors' contributions:
804	Conceptualization and design of the study: NA, CS, MP, KSL, SYB and AH; Data collection, data
805	analysis, investigation and interpretation: NA, BG, MP, AK, ALK, AS, SYB and CS; Writing –
806	original draft: NA and CS; Writing - review and editing: NA, CS, MP, ALK, KSL, AK, SYB, AH and
807	AS; Visualization: NA, CS and SYB; Supervision: CS and AK; Project administration: NA, CS and
808	AK; All authors have read and approved the final manuscript.

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- 815

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983		cells from primary and metastatic sites of epithelial ovarian cancers. Journal for
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985		

987 Figure legends

988 989	Figure 1. Patient-derived 3D microtumors (PDM) derived from primary OvCa tumor
990	specimen show high viability (A) Efficiency of isolating OvCa PDM from a total of $n = 19$ fresh
991	primary OvCa tumor tissues samples. PDM were isolated from $n = 14$ specimen with a success
992	rate of 73.7%. (B) Viability of OvCa PDM models. Exemplary 2D images from 3D projections of <i>n</i>
993	= 4 OvCa PDM models confirm high viability according to Calcein-AM (viable cells) and SYTOX™
994	Orange (dead cells) staining. (C) Percentage of viable and dead cells in OvCa PDM. Viability was
995	assessed by an image-based analysis (see SI Methods) in $n = 4$ OvCa PDM models shown in
996	(B). Data are shown as mean values with SEM from at least $n = 3$ PDM of each model. * $P < 0.05$,
997	** P < 0.01, *** P < 0.001, multiple paired t-test with Holm-Šídák's post hoc test. Scale bar 50 µm.

998 Figure 1. OvCa PDM resemble histopathological features of the corresponding primary

999 tumor tissue. Hematoxylin and Eosin (H&E) staining of OvCa PDM and corresponding primary

1000 tumor tissue (PTT) sections revealed features of malignant cells (incl. giant cells with more than

1001 one nucleolus, hyperchromatic cells with dark nuclei and high nuclei:cytoplasma ratio) confirming

1002 the cancerous origin. Qualitative characterization of OvCa PDM was performed by DAB

1003 immunohistochemical staining of FFPE sections (3 μm). Expression of OvCa histotype specific

1004 markers (p53, WT1), tumor markers (CA125, MSLN), tumor-associated macrophages (CD163),

 $1005 \qquad \text{immune/tumor marker (PD-L1), cancer-associated fibroblasts (FAP\alpha) and extracellular matrix}$

1006 components (Hyaloronan C1QBP, Collagen I) within PDMs is shown. Scale bars indicate 500 μm

1007 for PTT; 50 μ m for PDM; 20 μ m for magnifications (PTT and PDM). FAP α , cancer-associated

1008 fibroblast protein alpha; C1QBP, hyaluronan binding protein; WT1, wilms tumor 1; MSNL,

1009 mesothelin.

1010 Figure 3. RPPA protein profiling of OvCa PDM identifies significant differences in active

1011 protein signaling pathways as molecular basis for OvCa PDM drug treatment responses.

- 1012 (A) Protein heat map covering 116 analytes analyzed in OvCa PDM (n = 7) and BC PDM (n = 1)
- 1013 generated from sample sizes of n = 100-150 PDM. Protein abundances for each analyte are

displayed as median-centered, log₂-transformed NFI signals. Samples were subjected to

hierarchical clustering using Euclidean distance (complete linkage). (B) Activation state of

1014

1015 1016 different pathways in the different OvCa PDM models. Proteins related to an "active" pathway 1017 were selected for each of the plotted pathways (see Table S4). Protein signals are shown as 1018 median-centered, log₂ transformed NFI signals. Dotted lines indicate log₂ values of +0.6 (fold 1019 change of +1.5) and -1 \log_2 (fold change of -0.5). Data are shown as box and whiskers plots with 1020 minimum and maximum range. *P < 0.05, **P < 0.01, ***P < 0.001, Kruskal-Wallis test with 1021 Dunn's post hoc test. (C) Cytotoxicity measurement of indicated OvCa PDM models treated with 1022 standard platinum-based chemotherapy (Carboplatin; final concentrations 75, 100 and 125 µM). 1023 Four replicates per treatment with n = 15 PDM per well were performed and measured after 24 h, 1024 48 h and 72 h. Signals were measured as RFU (Relative Fluorescent Unit), background corrected 1025 and normalized to vehicle control (DMSO). Data are shown as mean values. Statistical significances compared to vehicle control or H₂O are shown. *P < 0.05, **P < 0.01, ***P < 0.001, 1026 1027 Two-way ANOVA with Bonferroni's multiple comparison test. Carbo: carboplatin 1028 Figure 4. Carboplatin drug response in OvCa PDM correlates with the activity of diverse 1029 signaling pathways. (A) Heat map of protein abundances (calculated from median-centered NFI 1030 values) averaged over carboplatin responder and non-responder OvCa PDM. Carboplatin 1031 responders and non-responders were grouped according to significant treatment effects from 1032 functional compound testing (Fig. 3C). Only proteins with >20% increased or decreased 1033 abundance between responder and non-responder group were selected. Data was HCL clustered 1034 with Euclidean distance (average linkage). (B) Signaling pathway activation in carboplatin 1035 responder vs. non-responder OvCa PDM. Proteins were sorted according to their pathway 1036 affiliation and according to upregulation or downregulation within responder group. *P < 0.05. **P1037 < 0.01, ***P < 0.001, Mann-Whitney-U-test. R, responder; Non-R, Non-responder. (C) Proteomic 1038 on- and off-target pathway effects in carboplatin-treated (75 µM) OvCa #24 PDM analyzed by 1039 RPPA. Treated PDM were analyzed from an immediate (0.5 h), an early (4 h) and a late (72 h) 1040 treatment time. For each time point, protein values are displayed as log₂-transformed treatment1041 to-control signal ratios (TR) calculated from NFI signals of treated PDM and corresponding

1042 vehicle control (DMSO). Only proteins with >50% differential protein abundance compared to

1043 vehicle control were selected. Straight lines above plots indicate statistical significances

1044 compared to vehicle control. *P < 0.05, **P < 0.01, ***P < 0.001, One-way ANOVA using

1045 nonparametric Kruskal-Wallis with Dunn's ad hoc test.

1046 Figure 5. CPI treatment in OvCa PDM-TIL co-cultures increased functional TIL killing

1047 **capacity.** Autologous TIL populations were isolated and expanded from OvCa tissue specimen.

- 1048 (A) Percentages of different TIL populations within CD3, CD8 and CD4 positive T cells of different
- 1049 models were quantified by multicolor flow cytometry. Data are shown as means ± SEM of at least

1050 n = 10 OvCa samples. **P* < 0.05, ***P* < 0.01, ****P* < 0.001, ANOVA with Holm-Šídák's post hoc

1051 test. (B) Phenotypes of extracted TIL populations shown separately for each OvCa model.

1052 Pembro: pembrolizumab 60 μg/ml; Ipilim: ipilimumab 50 μg/ml; Atezo: atezolizumab 50 μg/ml. (C)

1053 Percentages of CD8⁺ and CD8⁺CD39⁺ TILs in OvCa patients with lymph node spread (N = 1) and

1054 without lymph node spread (N = 0). All points with median are shown. *P < 0.05, **P < 0.01,

1055 Mann-Whitney-U-test. (D) PDM killing effects were measured in an image-based assay format as

1056 ratio of fluorescent intensities (FI) of dead cells vs. live PDM cells. Per treatment *n* = 3 PDM in

1057 three replicates were analyzed. Masks for viable PDM (Calcein-AM staining), dead cells

1058 (SYTOX[™] Orange dead cell staining) and TILs (CellTracker[™] Deep Red staining) were applied

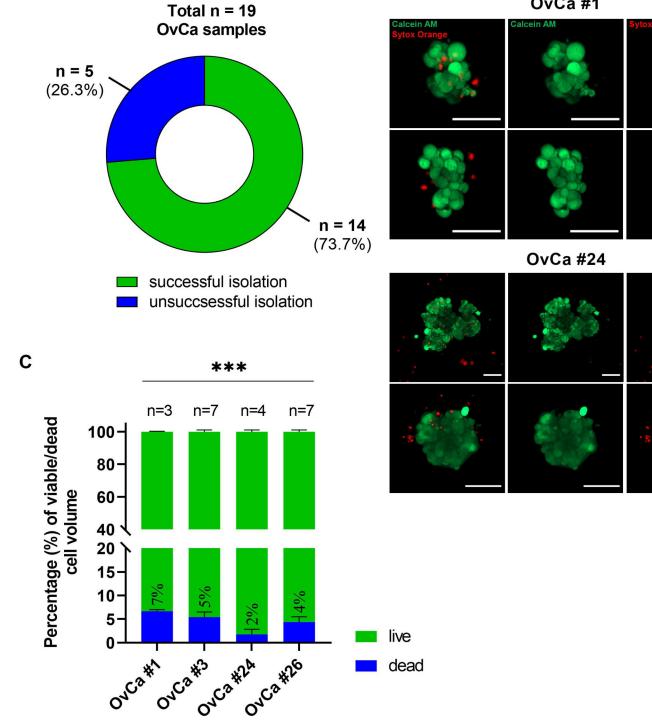
1059 using Imaris 8.0 software. Scale bars indicate 50 µm. FI from TILs were subtracted from the total

1060 dead FI. (E-F) Killing effects of autologous TILs on corresponding PDM in co-cultures treated with

immune checkpoint inhibitors (CPI). TILs of OvCa #24 (**D**) and #26 (**E**) were co-cultured with *n* =

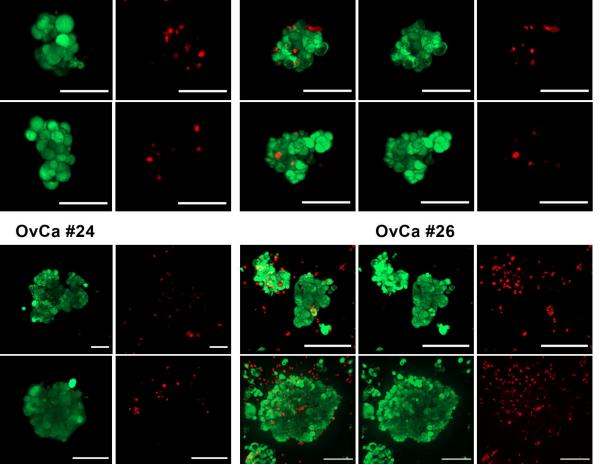
1062 15 PDM using an E:T ratio of 4:1 and treated with CPI either alone or in combination. *P < 0.05,

1063 ***P* < 0.01, ****P* < 0.001, ANOVA with Holm-Šídák's post hoc test.



В

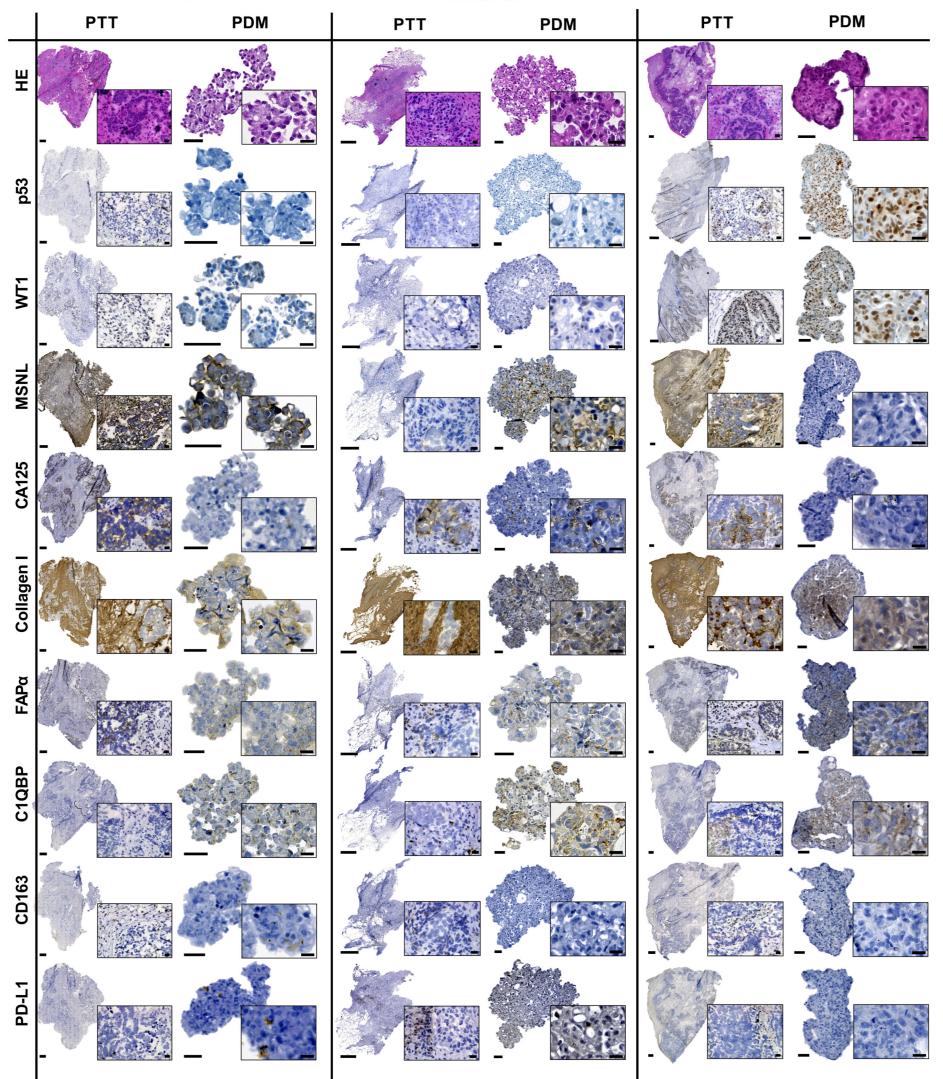
OvCa #1

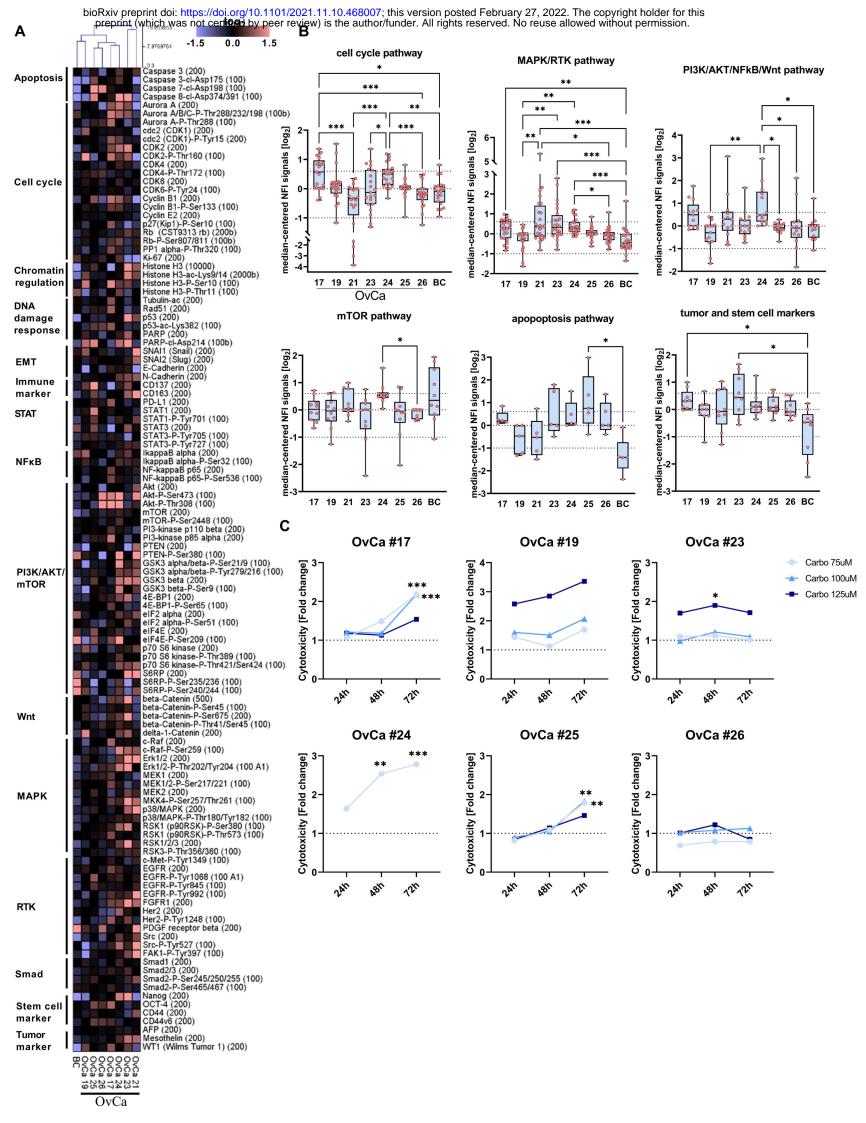


OvCa #3

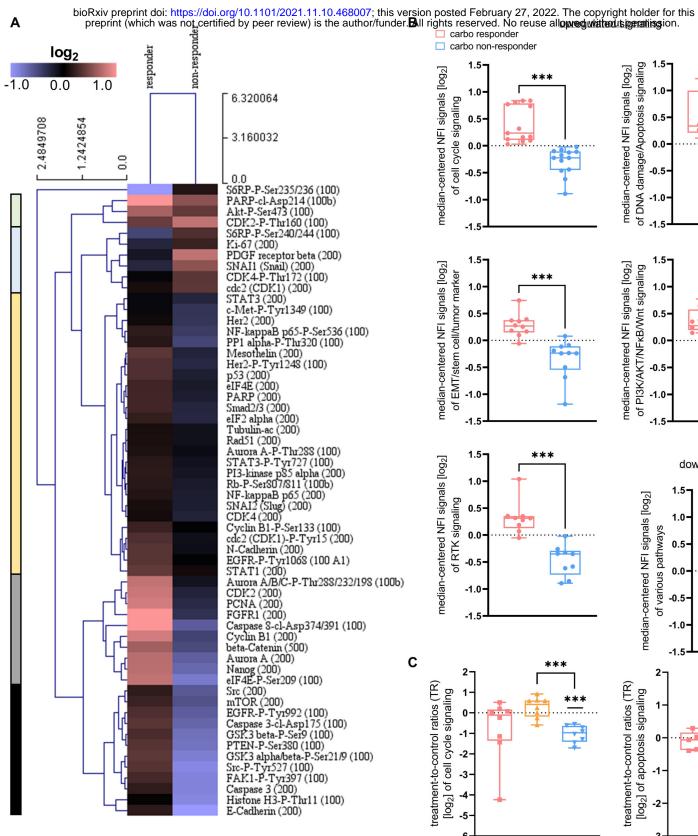
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OvCa #23

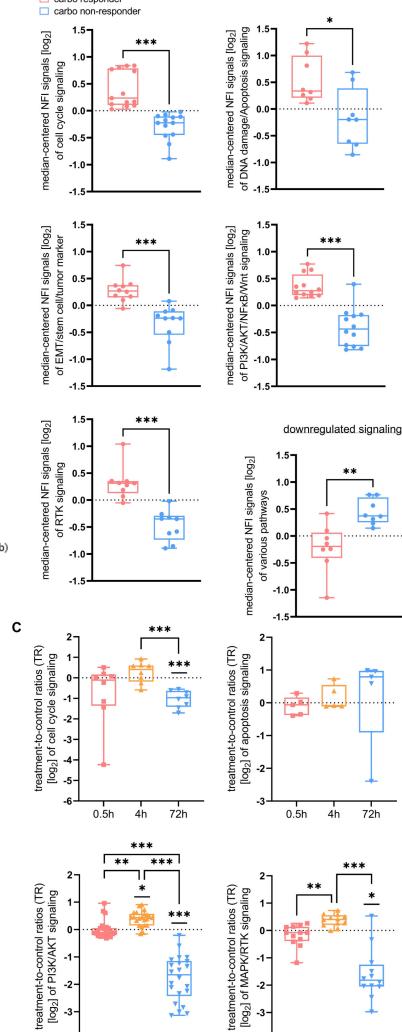




Α



High expression in R and Non-R High expression in Non-R Moderate expression in R High expression in R Low expression in Non-R



0.5h

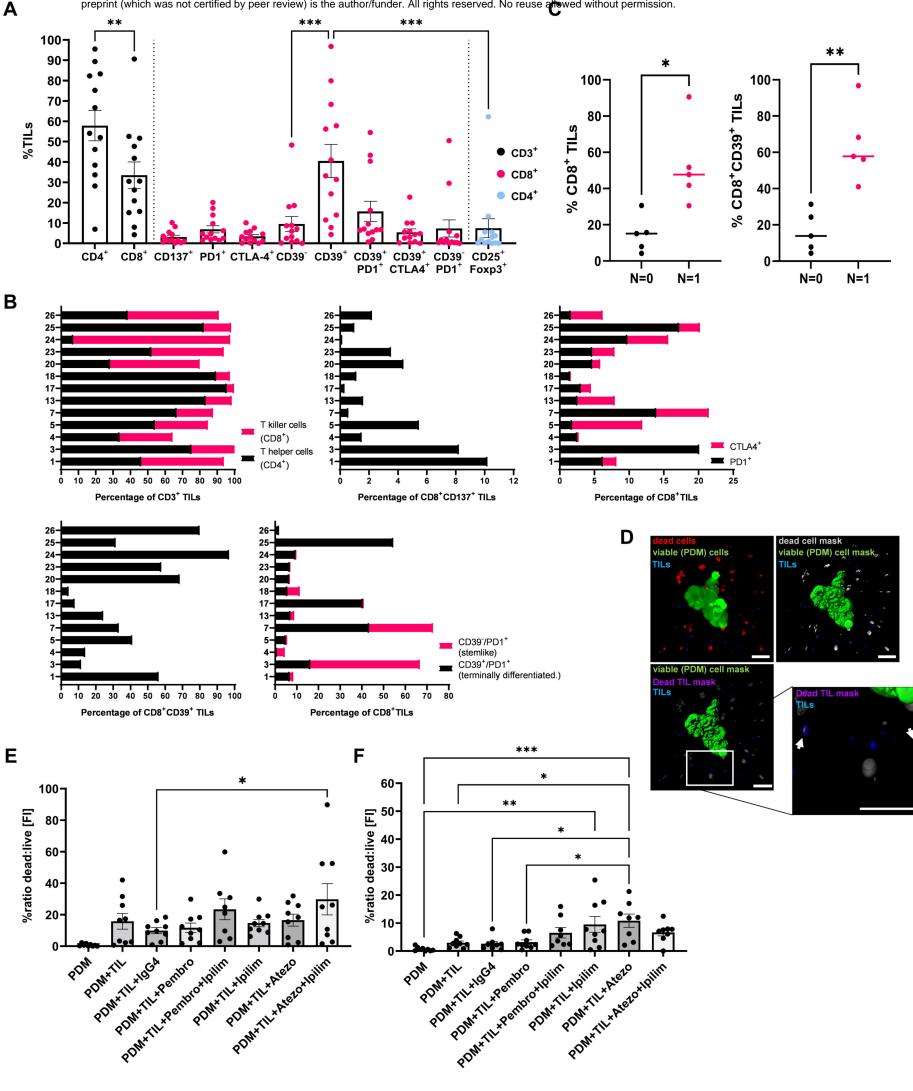
4h

72h

72h

0.5h

4h



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