

Title: COVID-19 during pregnancy alters circulating extracellular vesicle cargo and their effects on trophoblast

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One Sentence Summary: Circulating extracellular vesicles are altered by COVID-19 during pregnancy and contribute to placental dysfunction.

Abstract:

SARS-CoV-2 infection and the resulting coronavirus disease (COVID-19) complicate pregnancies as the result of placental dysfunction which increases the risk of adverse pregnancy outcomes. While abnormal placental pathology resulting from COVID-19 is common, direct infection of the placenta is rare. This suggests maternal response to infection is responsible for placental dysfunction. We hypothesized that maternal circulating extracellular vesicles (EVs) are altered by COVID-19 during pregnancy and contribute to placental dysfunction. To examine this, we characterized maternal circulating EVs from pregnancies complicated by COVID-19 and tested their functional effect on trophoblast cells *in vitro*. We found the timing of infection is a major determinant of the effect of COVID-19 on circulating EVs. Additionally, we found differentially expressed EV mRNA cargo in COVID-19 groups compared to Controls that regulates the differential gene expression induced by COVID-19 in the placenta. *In vitro* exposure of trophoblasts to EVs isolated from patients with an active infection, but not EVs isolated from Controls, reduced key trophoblast functions including hormone production and invasion. This demonstrates circulating EVs from subjects with an active infection disrupt vital trophoblast function. This study determined that COVID-19 has a long-lasting effect on circulating EVs and circulating EVs are likely to participate in the placental dysfunction induced by COVID-19.

INTRODUCTION

Maternal SARS-CoV-2 infection and the resulting coronavirus disease (COVID-19) is associated with an increased risk of pregnancy complications including preterm birth, hypertensive disorders of pregnancy, fetal growth restriction, and pregnancy loss (1-5). Placental dysfunction is known to contribute to these complications, and placental pathology, including vasculopathies and inflammation, is frequently reported following an acute or even resolved infection during pregnancy (6-11). This suggests COVID-19 has a long-lasting effect on maternal pregnancy via placental pathology. Despite extensive reports of placental pathology observations following COVID-19, little is known about the underlying mechanisms contributing to placental dysfunction and the potentially related, subsequent pregnancy complications. Direct infection of the placenta is rare which suggests placental dysfunction is caused by the maternal response to SARS-CoV-2 infection (10, 12, 13).

Circulating extracellular vesicles (EVs) are known to be altered by SARS-CoV-2 infection and contribute to COVID-19 induced organ damage and dysfunction.(14-19) EVs are a means of cell-to-cell communication because they carry bioactive cargo that elicits signaling events in recipient cells. Circulating EVs constitute a diverse population of vesicles carrying bioactive cargo that arise from many cellular sources. Two pathways of biogenesis create what is commonly referred to as large and small EVs (LEVs and SEVs). When compared with non-infected patients, EV cargo in patients with COVID-19 is altered, which elicits downstream effects such as coagulopathy (20-25) and inflammation (15, 16, 22, 23, 26, 27). Tissue factor abundance in EVs is also increased in COVID-19 and correlates with inflammation and disease severity (14, 17-19). These examples demonstrate the influence that circulating EVs have on the systemic response to COVID-19, thereby leading to distal organ dysfunction.

During normal pregnancy, the placenta releases EVs into the maternal circulation (28-30). Placental-derived EVs promote maternal adaptation to support a healthy pregnancy including a shift in the maternal immune system to a tolerant state (31, 32) as well as angiogenesis promotion (33). Placental-derived EVs also affect trophoblast function via autocrine and paracrine signaling. Trophoblasts are a specialized cell type of the placenta that are responsible for invasion into maternal tissue to anchor the placenta, vasculature remodeling for adequate placental blood flow, and hormone production for maternal and fetal signaling. Perturbation of normal trophoblast function, as caused by altered EV signaling, underlies the placental dysfunction that contributes to many pregnancy complications (34-41). Importantly, EV characteristics such as number and cargo are altered in multiple pregnancy complications (42).

We hypothesized that COVID-19 induces placental morphologic and transcriptomic changes and alters EV characteristics which have functional consequences in the placenta. Similar to previous studies, COVID-19 during pregnancy induced marked placental histologic and transcriptomic changes which were dependent on the gestational timing of COVID-19. Importantly, we found that COVID-19 altered EV cargo and trophoblast exposure to EVs resulted in marked dysfunction including reduced cell invasion and hormone production.

RESULTS

Pregnancy outcomes following active and resolved SARS-CoV-2 infection

Subjects were enrolled at the time of delivery between July 2020-August 2022. Controls had no known SARS-CoV-2 infection during pregnancy. COVID-19 cases were divided based on the gestational timing of infection and either had a resolved infection that occurred in the 1st, 2nd, or 3rd trimester (R1, R2, R3), or had an active infection at the time of delivery (AI) (Table 1).

There were no significant differences in maternal age between groups, but there were significantly more Black subjects with an infection in the third trimester (resolved or active) compared to Controls. Subjects with an infection in the third trimester also had a lower gestational age at delivery than Controls (Table 1). Despite testing positive for SARS-CoV-2 at the time of delivery, the AI group had a higher incidence of asymptomatic infection than those with a resolved infection ($p < 0.0001$).

Multiple studies have shown that SARS-CoV-2 infection during pregnancy is associated with adverse pregnancy outcomes. However, to date, no study has examined the relationship between the timing of infection and pregnancy complications in a single cohort. We found that the type of adverse outcome differed depending on the gestational timing of infection. The incidence of gestational hypertension was increased in R1, whereas spontaneous preterm birth was increased in R2. Preeclampsia and medically indicated preterm birth were increased in R3 and AI compared to Controls (Table 1).

Placental pathology in COVID-19

Abnormal placental pathology is commonly reported in patients with active and resolved SARS-CoV-2 infections (7), however, there has yet to be a comprehensive assessment of placentas following maternal infection at various gestational ages. Similar to previous reports (6-11), we found either maternal or fetal vascular malperfusion (MVM & FVM) lesions were increased by COVID-19 (Table 1). Interestingly, high-grade MVM and perivillous fibrin deposition was increased in subjects with a resolved infection that occurred in the second or third trimester, but not the first trimester or with an active infection. This suggests that it takes a substantial amount of time for the placenta to recover from COVID-19. Alternatively, it is possible that the first trimester placenta is more resilient. The lack of high-grade MVM and

perivillous fibrin deposition in placenta collected from subjects with an active infection suggests these particular lesions take many weeks to manifest.

Timing of COVID-19 impacts the placental transcriptome

To gain insight into potential novel pathways that might be disrupted by COVID-19 infection during pregnancy, we sequenced the transcriptome of the placenta. Multiple genes were differentially expressed in COVID-19 groups compared to Controls (Supplemental Table 1). Not surprisingly, the transcriptome was most altered in placenta from pregnancies complicated by an active infection (Figure 1). When comparing the Control placenta to 3rd trimester resolved infection and active infection, DEGs were mostly downregulated (Supplemental Table 1). Interestingly, prostaglandin E receptor 2 (PTGER2) is decreased in Control placenta compared to AI. PTGER2 encodes for a prostaglandin receptor and downstream signaling regulates trophoblast behavior. In contrast, differentially expressed genes were predominately upregulated in Control placentas compared to R1. One of the few differentially expressed genes upregulated in R1 placenta encodes for a mitochondrial protein (*MTARC2*). Finally, there were a few differences between the R2 and control placentas. Interestingly there was overlap in DEGs in placentas in the resolved infection groups and the actively infected group (Supplemental Table 1). This suggests that COVID-19 induces permanent changes in gene expression in the placenta even when the infection is remote to delivery.

We used Ingenuity Pathway Analysis (IPA) to identify disrupted canonical pathways (Table 2). Several common pathways were altered in placenta from COVID-19 groups, including fibrosis and many inflammatory and immune regulating pathways. Thus, despite the long period of time following the resolution of maternal SARS-CoV-2 infection, the placenta has evidence of persistent inflammation and fibrosis.

Sustained effects on circulating EVs following COVID-19 in early pregnancy

EVs mediate cellular communication and play a fundamental role in the maintenance of tissue homeostasis and pathogenesis in several diseases. As discussed above, SARS-CoV-2 rarely infects the placenta and maternal circulating EVs are likely to play a role in mediating placental dysfunction. Therefore, we characterized EVs isolated from maternal plasma collected at delivery to determine if the EV profile was altered by COVID-19. The presence of large (LEV) and small EVs (SEV) in isolated particles was confirmed by morphological assessment of electron microscopy images (Figure 2A). Isolated LEVs and SEVs also had the expected size distribution of EVs (Figure 2B). Finally, the expression of the EV related tetraspanin, Cd9, was abundant in isolated particles (Figure 2C).

EV characteristics, including the concentration and size distribution, revealed persistent alterations in patients with a resolved infection. We found the diameter of SEVs was significantly increased in resolved infections compared to Controls, 108.7 vs. 117.2nm ($p=0.023$), but the difference in concentration was not significantly different, 2.15×10^8 vs. 1.56×10^8 EVs/uL plasma ($p=0.074$). When resolved infections were further divided by trimester, we found SEVs isolated from R2 circulation have an increased diameter and reduced concentration, but SEVs isolated from R1 and R3 were not different from Controls (Figure 3 C&D). In LEVs isolated from R2, we also found a decrease in their diameter but no change in the number of LEVs (Figure 3B). The uniquely persistent effect of COVID-19 on the concentration and size of circulating EVs in the second trimester may point to a particularly susceptible pregnancy-related pathway in this gestational window.

Altered cell of origin of EVs in patients with COVID-19 in early pregnancy

Characterizing the source of circulating EVs provides biological information about both the tissue and cell-type of origin as well as its functional state. We therefore used flow cytometry to detect cell-specific vesicle membrane protein expression and identified the relative contribution of each EV tissue-cellular source. The studied cells included those of maternal endothelial cells (Cd31+ Cd34-), fetal endothelial cells (Cd31+ Cd34+), platelets (Cd41a+), immune cells (Cd45+), and trophoblasts (PLAP+) (Figure 4).

Trophoblast derived PLAP+ EVs comprised the largest proportion of circulating LEVs (Figure 4A). The percentage of PLAP+ EVs was increased in the circulation of R2 compared to Controls suggesting the placenta secreted more LEVs into circulation. Interestingly, we found a subset of endothelial-derived EVs that also express Cd34, suggesting that these EVs originated from fetal endothelial cells (43). Fetal endothelial cell-derived LEVs (Cd34+ Cd31+) were also elevated in R1 and R2 compared to Controls. Surprisingly, the percentage of SEVs from the placenta was not altered by COVID-19. In fact, there was no difference in the percentage of SEVs from any cell type measured. This suggests that placenta-derived LEVs, but not SEVs, were persistently altered by COVID-19 in early pregnancy.

EVs alter trophoblast function *in vitro*

The placenta is made up of two main functional cell types- syncytiotrophoblasts, which are responsible for hormone production, and extravillous trophoblasts, which invade deep into the uterus to anchor the placenta and enable blood and nutrient flow to the fetus. EV signaling is known to influence trophoblast function (44). We tested the capacity for circulating EVs isolated from subjects with an active infection or Controls to alter the function of both major trophoblast cell types. To study syncytiotrophoblasts, we used the BeWo choriocarcinoma cell line which is an established cell line commonly used to study syncytiotrophoblast behavior.

After the addition of the cAMP producer forskolin, BeWo cells mimic syncytiotrophoblasts morphologically via syncytialization and functionally through the production of hormones such as hCG and progesterone. To assess extravillous trophoblasts (EVT) behavior, we used primary EVTs isolated from first trimester placenta and quantified the extent of invasion using an invasion assay. We chose to focus our *in vitro* experiments on EVs isolated from subjects with an active infection compared to Controls because the changes in placental pathology and the number of differentially expressed genes in the placenta was highest in AI cases compared to Controls.

Syncytiotrophoblast hormone production is vital to maintain a healthy pregnancy. The ratio of hCG to progesterone in the media was reduced following exposure to AI EVs compared to Control EVs (Figure 5A). A reduction of the ratio of hCG to progesterone indicates specific pathways related to hormone production were disrupted instead of reduced cellular activity which would result in an overall decrease in both levels and therefore a similar ratio. Altered pathways may include reduced transcription or translation which are key to hCG production or altered mitochondrial function which is key to progesterone synthesis. This demonstrates that AI EVs, but not Control EVs, disrupt trophoblast hormone production which may have profound effects on pregnancy maintenance.

Remarkably, extravillous trophoblast invasion was also reduced by exposure to AI EVs compared to Control EVs (Figure 5B). EVT invasion is vital to anchor the placenta to the maternal uterus and failure to appropriately do so increases the risk of preeclampsia, intrauterine growth restriction, and fetal loss.(45, 46) This extraordinary finding suggests that circulating EVs from pregnancies complicated by an active infection have the potential to directly reduce trophoblast invasion.

To identify novel pathways that may contribute to trophoblast dysfunction, we measured the BeWo transcriptome following EV exposure. AI EVs significantly altered gene expression in BeWo cells compared to Control EVs. 65 DEGs were identified, and the top differentially expressed genes encode for histone proteins (Table 3). This demonstrates a disruption of DNA packaging and transcription in trophoblasts exposed to AI EVs compared to Control EVs. Moreover, a top canonical pathway identified by IPA is related to cellular transcription (*Transcriptional Regulatory Network in Embryonic Stem Cells*) (Table 4) (47).

EV RNA cargo was altered in patients who experienced COVID-19 during pregnancy

Prior studies have shown that EVs carry bioactive cargo. Here, we sought to determine the cellular signaling effect of EV cargo on trophoblasts. Nucleic acids are the most commonly studied EV cargo, and small RNAs, particularly micro RNAs, are the most abundant (48). However, longer RNAs, such as messenger RNAs (mRNAs), have also been found and have not been widely studied. mRNAs encapsulated within EVs have been shown to be transferred to recipient cells and translated into proteins, altering the behavior of the cells (49-57). Therefore, we profiled the mRNA cargo content in circulating SEVs and LEVs to determine if there were differences dependent on the gestational timing of COVID-19. We sequenced an average of 2,946 transcripts in LEVs and 1,947 transcripts in SEVs.

We identified multiple uniquely expressed genes in EVs between the groups. Transcripts expressed by LEVs isolated from COVID-19 groups but absent in Controls include *YYIAP1*, *MOSPDI*, *RYBP*, and *HI-4* (Table 5). The proteins these transcripts encode are related to transcription, except for *HI-4* which has an unknown cellular function. In SEVs, *MYL4* and *C18orf32*, *CAPG*, and *CTSS* COVID-19 groups, but not in Controls (Table 6). These transcripts encode for motor and immune proteins. Many other genes were uniquely detected in LEVs and

SEVs isolated from individual COVID-19 groups (Table 5&6). The unique expression of these transcripts suggests that COVID-19 induces upregulation of these transcripts making them more available for EV packaging or increased specific transcript loading into EVs.

EVs isolated from Controls carry several interesting transcripts that were uniquely expressed in Controls, i.e. not in COVID-19 groups. LEVs isolated from COVID-19 groups are missing *APBA3*, *MTSS*, *FCF1*, *PSG2*, *LOC100128233*, *PHOSPHO2* and *THOC3* which are abundant in Control EVs (Figure 6A). These transcripts encode for proteins that were involved in various cellular functions. In contrast, only a few transcripts were unique to Controls in SEVs. These include *PYCR2*, *SCGB1C1*, and *CD300L4* (Figure 6B). *PYCR2* encodes for a cellular metabolism protein, *CD300L4* encodes for an immune regulating protein, and the protein function of *SCGB1C1* is unknown. Numerous other abundant Control transcripts were absent in individual COVID-19 groups and the corresponding expression in the other COVID-19 groups was commonly downregulated (Figure 6). The loss of these physiologically expressed genes in EVs isolated from COVID-19 groups suggest COVID-19 disrupts normal EV cargo.

The transcripts carried by EVs reflect the activity of the secreting cell. While the individual genes identified in LEVs and SEVs differ, the cellular functions of these genes often overlapped (Table 5&6, Figure 6). For example, large and small EVs isolated from R2 carried transcripts that control cell signaling, gene expression, immune regulation, and metabolism. Other overlapping cellular activities of interest were proliferation, apoptosis, invasion, ubiquitination, platelet function, and vesicle formation.

Many transcripts that were identified in LEVs and SEVs have been shown to be highly expressed in trophoblasts and previously reported to increase with gestation (Table 5&6 and Figure 6). Interestingly, LEV expression of *THOC3*, a highly expressed transcript in

trophoblasts, was abundant in Control LEVs but lowly expressed or absent in the COVID-19 groups. In addition, LEVs carry a different highly expressed trophoblast transcript, *RYBP*, in COVID-19 groups, but this transcript is absent in Controls. Both genes encode for transcription regulating proteins. Several of the uniquely expressed genes in EVs isolated from COVID-19 groups have been previously reported to be associated with adverse pregnancy outcomes, including fetal growth restriction, preeclampsia, gestational hypertension, and preterm birth (Supplemental Table 2). The number of these genes was highest in AI EVs, but also abundant in EVs isolated from subjects with resolved infections (Figure 7). Interestingly, eleven transcripts that have been implicated in preeclampsia were differentially expressed by AI EVs compared to Control EVs. Moreover, two transcripts associated with preterm birth were uniquely carried by R2 but not Control EVs. This suggests that there may be a shared etiology of these pregnancy complications because the incidence of these complications was also increased in these COVID-19 groups.

Regulators of COVID-19 induced differential gene expression in the placenta are in circulating EVs

We used IPA to identify upstream regulators of gene expression that were carried by circulating EVs in COVID-19 groups and determined whether their downstream target genes were differentially expressed in placenta. Surprisingly, we found many upstream regulators were carried by circulating EVs and were differentially expressed between the COVID-19 group and the Controls (Table 7). For example, Jun proto-oncogene B (*JUNB*) mRNA levels were increased in LEVs from R1 compared to Controls (FC=-2.25, p=0.01) and JUN signaling was altered in R1 placenta compared to Controls. Likewise, interleukin 1 beta (*IL1 β*) mRNA was decreased in R3 LEVs compared to Controls (FC=1.85, p=0.07) and IL1 β signaling was also

altered in R3 placenta compared to Controls. Finally, estrogen receptor 1 (*ESR1*) mRNA was decreased in R3 SEVs compared to controls (FC=6.99, p=0.13) and ESR1 signaling was predicted to regulate differential gene expression in R3 placenta compared to Controls. These examples exemplify potential EV driven signaling that cause differential gene expression in the placenta.

mtDNA was abundant in LEVs and increased following COVID-19 in early pregnancy

Analysis of the placental transcriptome following COVID-19 identified differentially expressed genes indicative of mitochondrial dysfunction. To determine if EVs were enriched in mitochondrial cargo, we measured mitochondrial DNA (mtDNA) in circulating EVs (Figure 8A &D) and found mtDNA was more abundant in large than small EVs. Also, the abundance of mtDNA in LEVs, but not SEVs, inversely correlated with the gestational timing of infection (Figure 8 B&D). This suggests that the mitochondrial function of cells producing LEVs were disrupted by COVID-19.

DISCUSSION

The long-term effects of COVID-19 during pregnancy have not yet been elucidated. Previous studies, and what is reported here, show that the placenta is persistently damaged, and the likelihood of adverse pregnancy outcomes increases in patients with a pregnancy complicated by SARS-CoV-2 infection. Here, we have begun to dissect out the mechanisms underlying the observed placenta pathology and dysfunction. Importantly for the first time, we demonstrate that circulating EVs from COVID-19 affected pregnancies 1) have a detrimental effect on trophoblast function, including hormone production and invasion, *in vitro*; 2) are persistently altered after SARS-CoV-2 infection; and 3) carry cargo that has been previously associated with adverse pregnancy outcomes.

Trophoblast dysfunction following exposure to EVs isolated from subjects with an active SARS-CoV-2 infection *in vitro* provides evidence that circulating EVs contribute to the resulting placental pathology. We chose to focus our *in vitro* trophoblast experiments on the response to EVs isolated from subjects with an active infection because the magnitude of alteration to the placental transcriptome was greater compared to those placentas from resolved infections. Analysis of the BeWo transcriptome after AI EV exposure showed that top differentially expressed genes encode histone proteins, suggesting alterations in chromatin packaging. Disrupted DNA replication may be the underlying mechanism leading to reduced hormone production following the same exposure. Trophoblast hormone production is key to maintaining a healthy pregnancy and altered levels are associated with pregnancy complications such as preeclampsia and fetal demise *in utero* (58). Reduced BeWo hormone production following exposure to AI EVs suggests a failure of the placenta to produce hormones in the presence of circulating EVs. Extravillous trophoblast invasion is key to implantation, and failure to appropriately invade is implicated in the pathogenesis of preeclampsia, fetal growth restriction, and preterm birth (45, 46). Therefore, the reduction of EVT invasion following exposure to AI EVs, suggests that EVs have the capacity to reduce invasion and impede placentation. Collectively, the effect of circulating EVs isolated from patients with an active infection on trophoblasts *in vitro* are profound and suggest AI EVs have the capacity to contribute to the placental pathology observed in pregnancies complicated by COVID-19.

Gestational age at the time of infection is a major determinant of COVID-19-induced changes in the profile of circulating EVs. If infected during the first or second trimester of pregnancy, trophoblast and fetal endothelial cell LEV numbers are increased and circulating LEVs carry more mtDNA. This suggests that mitochondrial function in the placenta in early

pregnancy is perturbed by COVID-19. Mitochondrial dysfunction is also reported in many organs following SARS-CoV-2 infection and is thought to contribute to cell death, inflammation, and cell injury (59-63). Appelman et al. recently reported persistent mitochondrial dysfunction in skeletal muscle long after the resolution of SARS-CoV-2 infection (64). Cells undergoing mitochondrial dysfunction release mtDNA. In support of this, Faizan et al. recently demonstrated that SARS-CoV-2 infected airway epithelial cells with mitochondrial dysfunction release mtDNA in EVs (60). Further, elevated circulating mtDNA, is well established in COVID-19 and correlates with severity and length of infection (65-68). Thus, our results suggest that abnormal mitochondria may also play a role in the pathogenesis of placental dysfunction in COVID-19 as we and others have shown.

Circulating EV cargo reflects the activity of the cells of origin. The transcripts carried by EVs encode genes related to inflammation, vasculopathies, bioenergetics, and cell death. These pathways are also evident in the transcriptome and histopathology of the placenta regardless of the timing of infection. Many of the transcripts carried by EVs are highly expressed by trophoblasts and have known roles in trophoblast function including cellular metabolism, transcription, and immune regulation. We also found that many of the transcripts in EVs from pregnancies complicated by COVID-19 are also implicated in adverse pregnancy outcomes including preterm birth, fetal growth restriction, gestational hypertension, and preeclampsia. This points to shared pathways of placental dysfunction induced by a distal SARS-CoV-2 infection.

EV cargo is selectively packaged and functionally active upon interaction with recipient cells. RNA is selectively incorporated into EVs and is biologically active in target cells (48). Mitochondrial DNA packaged in EVs can alter the recipient cell's mitochondrial function (69-

71). The capacity for EVs to elicit a functional response in recipient cells is illustrated by the trophoblast response to EVs isolated from patients with an active infection.

The overlap of uniquely expressed genes in EVs and the IPA predicted upstream regulators of the placental transcriptome suggests EV mRNA cargo may participate in altering the placental transcriptome. JUN signaling is altered in R1 placenta compared to Control placenta and *JUNB* mRNA is increased in Control LEVs compared to R1 LEVs. JUN proteins are important for placentation, and a loss of JUN signaling is implicated in preeclampsia (72, 73). Thus, low levels of *JUNB* in COVID-19 EVs may suggest placental dysfunction leading to the later development of preeclampsia, which is observed at higher rates in pregnancies complicated by COVID-19 (74). In R3, hormone receptor signaling is identified as a top canonical pathway and differentially expressed genes are regulated by *ESR1*. *ESR1* mRNA is abundant in SEVs isolated from Controls but not R3. *ESR1* and other hormone receptor signaling is important for maintaining a healthy pregnancy and placental function. The presence of upstream regulators in circulating EVs suggest placental dysfunction may in fact be a result of EV cargo delivery.

Our study is limited by the number of symptomatic patients with an active infection at the time of delivery. Despite only 14% of patients experiencing COVID-19 related symptoms, the placenta had significant pathology and altered transcriptome, and there was an increase in the incidence of preeclampsia and medically indicated preterm birth in asymptomatic and symptomatic AI cases. Interestingly, EVs isolated from asymptomatic patients also had profound effects on trophoblasts *in vitro*. This suggests that we are underestimating the effect of a symptomatic active SARS-CoV-2 infection on circulating EVs and placental function.

Our study demonstrates the effect of maternal SARS-CoV-2 infection on circulating EVs and that EVs can contribute to placental pathology following COVID-19. This is the first study to analyze the effect of maternal circulating EVs on trophoblast function. The placental transcriptome and EV cargo reveal overlap in pathways that underlie pregnancy complications caused by maternal COVID-19 and other less well understood disorders. Future studies should test if the maternal circulating EVs in other pregnancy disorders contribute to placental dysfunction.

Figure 1. Differential gene expression in the placenta induced by COVID-19 during pregnancy. The number and direction of differentially expressed genes between Controls and resolved infections in the first trimester (R1), second trimester (R2), and third trimester (R3) and active infection (AI). (n=9-10/group)

Figure 2. Extracellular vesicles are isolated from the plasma by serial centrifugation. **(A)** Representative transmission electron microscope images demonstrated EV structure including a lipid membrane. **(B)** Nanotracking analysis was used to determine the size of particles in three control samples **(C)** Cd9 expression evaluated by immune-blotting was abundant in isolated particles in three control samples.

Figure 3. Circulating EVs were persistently altered in patients who experienced COVID-19 in the second trimester. **(A)** The number of LEVs in the circulation at the time of delivery (n=14-22/group). **(B)** The diameter of LEVs in the circulation (n=14-22/group). **(C)** The number of SEVs in the circulation at the time of delivery (n=14-22/group). **(D)** The diameter of SEVs in the circulation (n=14-22/group). All data are presented as mean \pm SD. All analyses were performed by one-way ANOVA or the Kruskal-Wallis test, followed by post-hoc tests. Comparisons were made between Controls and resolved infection in the 1st trimester (R1), second trimester (R2), third trimester (R3), and active infection (AI).

Figure 4. COVID-19 in early to mid-pregnancy increased the contribution of endothelial cells and trophoblasts to the circulating LEVs. **(A)** Relative frequency of LEVs derived from endothelial cells, platelets, immune cells, and trophoblasts (n=11-16). **(B)** The relative frequency of SEVs derived from endothelial cells, platelets, immune cells, and trophoblasts (n=13-20/group). All data are represented as mean \pm SD. All analyses were performed by one-way ANOVA or Kruskal-Wallis test, followed by post-hoc tests. Comparisons were made between

Controls and resolved infection in the 1st trimester (R1), second trimester (R2), third trimester (R3), and active infection (AI).

Figure 5. Trophoblast function was altered by exposure to EVs isolated from patients with an active infection (AI) compared to Controls. **(A)** Human chorionic gonadotropin (hCG) and progesterone were measured in the media of forskolin-treated (syncytialized) BeWo cells. The ratio of hCG to progesterone was normalized to hormone production by cells not exposed to EVs (noEVs). The results of two experiments (identified by symbol shape) are reported in A. (n=6-7/group, 2 experiments). **(B)** Extravillous trophoblasts (EVTs) were isolated from three placentae (identified by symbol shape) and exposed to Control or AI EVs (n=9-10/group, 3 experiments). Invasion was calculated and normalized to invasion of EVT not exposed to EVs. All data are presented as mean \pm SD. All analyses were performed by one-way ANOVA or Kruskal-Wallis test, followed by post-hoc tests.

Figure 6. Control EVs carried transcripts that were absent in COVID-19 groups. **(A&B)** Genes were identified as uniquely expressed in Controls if all comparison samples had no expression (gray bars). The expression of each COVID-19 group compared to Controls is shown in the heatmap. The lower the expression in COVID-19 groups the darker the blue, and if expression was increased in a COVID-19 group compared to Controls it was colored red. The general cellular function of each gene is listed on the right. LEV transcripts are reported in (A) and SEV transcripts are reported in (B) (n=9-10/group). Comparisons were made between Controls and resolved infection in the 1st trimester (R1), second trimester (R2), third trimester (R3), and active infection (AI). Transcripts known to be highly expressed by trophoblasts are marked by (T).

Figure 7. Uniquely expressed genes associated with adverse pregnancy outcomes were expressed in EVs. The number of genes with previously reported altered expression in fetal

growth restriction (FGR), preeclampsia (PE), gestational hypertension (gHTN), and preterm birth (PTB) are quantified.

Figure 8. LEV abundance of mtDNA is inversely correlated with gestational timing of infection.

(A&C) The amount of mtDNA in each LEV (A) and SEV (C) is reported for each group (n=13-20/group). Data are presented as mean \pm SD and tested by ANOVA followed by post-hoc tests.

Comparisons were made between Controls and resolved infection in the 1st trimester (R1),

second trimester (R2), third trimester (R3), and active infection (AI). **(B&D)** The Pearson

correlation between gestational age at infection and mtDNA content is reported for LEV (B) and

SEV **(D)**.

MATERIALS AND METHODS

Patient Cohort

The COMET study was conducted at the Hospital of the University of Pennsylvania (HUP) with Institutional Review Board approval (IRB#843277). Subjects received a description of the study and signed an informed consent before enrollment. Subjects were enrolled at the time of delivery in the COMET study between April 2020-June 2022. Subjects were tested for a SARS-CoV-2 infection by nasopharyngeal polymerase chain reaction (PCR) upon admission to the labor and delivery unit at HUP. Subjects who tested positive at the time of delivery were enrolled in the active infection (AI) group. Those subjects who tested negative and had no known SARS-CoV-2 infection during their pregnancy were included as Controls. Subjects with a negative test at delivery and a history of SARS-CoV-2 infection, during their pregnancy and greater than 14 days before enrollment, were included in the resolved infection group (R). The gestational age of SARS-CoV-2 infection was calculated, and subjects were further divided into the trimester of

infection (resolved infection in the first trimester (R1), resolved infection in the second trimester (R2) and resolved infection in the third trimester (R3)).

Clinical and Demographic Data Collection

Clinical characteristics, such as maternal age, self-identified race, gestational age at infection, and pregnancy outcomes, were extracted from the medical record (Table 1).

The severity of COVID-19 disease was categorized based on the National Institute of Health and Society for Maternal-Fetal Medicine definitions: Asymptomatic infection are individuals who test positive but experience no symptoms. Symptomatic patients include all levels of illness (mild-critical).

Sample Collection

Placenta were collected at the time of delivery. All placentas were examined by the pathology department at the Hospital of the University of Pennsylvania (HUP). Placenta were reviewed following a systematic protocol which includes recording the trimmed placental weight, membrane insertion site, gross appearance, dimensions of the placental disc, and umbilical cord insertion, length, and diameter. Full thickness placental biopsies were collected and fixed in 10% formalin for histological assessment. Macroscopic and microscopic lesions were identified and classified according to the Amsterdam Placental Workshop Group 2014 classification (75-77). Placental biopsies were also collected and stored in Trizol for RNA sequencing.

Blood was collected at the time of delivery in an EDTA tube and spun at 1,000G for 10 minutes at room temperature to isolate plasma, which was aliquoted and stored at -80.

Placenta and BeWo RNA Isolation and Sequencing

Total RNA was isolated from placenta samples using Qiagen RNEasy Plus Mini Kits (Cat# 74134 Qiagen, Hilden, Germany). Total RNA was isolated from BeWo cells with the PicoPure RNA Isolation Kit (Cat# KIT0204 Applied Biosystems, Waltham, MA). Isolated RNA was sent to NovoGene for library preparation and sequencing.

RNA integrity and quantification was assessed using the RNA Nano 6000 Assay Kit of the Bioanalyzer 2100 System (Agilent Technologies, CA, USA). RNA purity was checked using a NanoPhotometer spectrophotometer (IMPLEN, CA, USA). A total of 1 μ g RNA per samples was used as input material for the RNA sample preparation. Sequencing libraries were generated using NEBNext Ultra RNA Library Prep Kit for Illumina (NEB, USA) following manufacturer recommendations, and index codes were added to identify samples. Clustering of the index-coded samples was performed on an Illumina Novaseq sequencer according to manufacturer's instructions. After cluster generation, libraries were sequenced, and pair-end reads were generated. Raw data (raw reads) of FASTQ format were processed through fastp. and clean data was obtained by removing reads containing adapter and poly-N sequences and reads with low quality. Pair-end clean reads were aligned to the GRCh38/hg38 reference genome using Spliced Transcripts Alignment to a Reference (STAR) software. FeatureCounts were used to count the read number mapped to each gene. Then RPKM of each gene was calculated based on the length of the gene and read count mapped to the gene. Differential gene expression between COVID-19 groups and Controls was tested by DESeq2 method. Differentially expressed genes were determined based on their adjusted p-value (<0.05) and >1.5-fold change. Functional analysis was conducted using Qiagen's ingenuity pathway analysis (IPA). Canonical pathways and upstream regulators with a p-value <0.05 were considered.

EV Isolation and Characterization

Serial centrifugation was utilized to isolate EVs from plasma. One mL of plasma was spun in an Eppendorf 5424 benchtop centrifuge at 2,000 x g for 10 minutes at 4°C. The supernatant was then spun at 20,000 x g for 30 minutes at 4°C. The pellet (LEVs) was washed in 1mL filtered PBS and spun again at 20,000 x g for 30 minutes at 4°C. The LEV pellet was resuspended in 100uL filtered PBS. The supernatant was spun by the Beckman Ultracentrifuge Optima Max TL using the TLA 120.2 rotor at 100,000 x g (48,000RPM) for 90 minutes at 4°C. The pellet (SEVs) was washed with 1mL filtered PBS and spun again at 100,000 x g (48,000RPM) for 90 minutes at 4°C. The SEV pellet was resuspended in 100uL filtered PBS.

EV isolation was confirmed by transmission electron microscopy, nanoparticle tracking, and protein measurement. Transmission electron microscopy images were generated and resulting images reviewed for the presence of EVs. EVs were analyzed by Particle Metrix Zetaview nanoparticle tracking. 11 fields were captured using the following parameters: sensitivity 80, frame rate 30, shutter 100, minimum brightness 1000, minimum area 10, trace length 15. Representative histograms and TEM images for LEV and SEV are included in Figure 2. Cd9 protein abundance was determined by gel electrophoresis. Total protein was measured with a Qubit Protein Assay Kit and EV suspension was evaporated by vacuum and resuspended in electrophoresis buffer. Three LEV (5µg) and SEV (20µg) samples were loaded into BioRad Mini-Protean TGX Gel 4-20% polyacrylamide gels with Licor Chameleon Duo ladders (928-60000) and run at 20mA for 2 hours. Proteins were transferred to nitrocellulose membrane via 200mA over 3 hours on ice. The membrane was blocked in Licor Intercept Blocking Buffer for 1 hour at room temperature then incubated with Cd9 antibody (HI9a Biolegend Cat 312112) at 1:5000 overnight at room temperature. The membrane was then incubated with Licor IRDye

800CW streptavidin (926-32230) at 1:5000 for 2 hours at room temperature and the membrane was imaged by Licor Odyssey.

Flow Cytometry on EVs

EVs surface protein expression was determined by flow cytometry. EVs were resuspended at $1 \times 10^8 / 10 \mu\text{L}$ of filtered PBS. Di-8-ANEPPS (Invitrogen Cat# D3167) was reconstituted in ethanol as per manufacturer instruction and further diluted to 1:1000 in filtered PBS. Antibodies were spun at 20,000 x g for 30 minutes at 4°C immediately before use. 10 μL of EV suspension was incubated in ANEPPS (10 μL) and antibodies, 1.25 μL Cd45-Ry586 (Cat# BD568135), 1.25 μL Cd41a- PE/Cy7 (Cat# BDB561424), 1.25 μL PLAP- eFlour660 (Fisher Cat# 50-112-4573) , and 1.25 μL Cd34- PE/CF594 (Cat# BDB562449) and 3 μL Cd31-AF700 (Biollegend Cat# 50-207-2950), for 30 minutes at room temperature. 470 μL of filtered PBS was added before samples were measured by BD Symphony A1 cytometer. Negative controls included: antibodies alone, EVs without ANEPPS, and EVs treated with 1% triton. Data was analyzed using FlowJo software. EVs were identified by Small Particle Side Scatter (SP-SSC) and expression of ANEPPS.

***In vitro* trophoblast EV co-culture**

BeWo cells, subclone B30, were cultured in 75-cm² flasks (Fisher Scientific) at 37°C and 5% CO₂ in media (DMEM/F12, 10% FBS, 1% P/S, 1% L-alanyl-L-glutamine). EV-depleted media was made with EV-depleted FBS (Gibco A2720801) and used for cell culture experiments. Cells were plated at 250,000 cells/well in a 6 well plate and 1.5mL of EV-depleted media was added. Cells adhered for 24 hours before adding 1 $\mu\text{g}/\mu\text{L}$ forskolin, cAMP producer, to promote syncytialization for an additional 24 hours. LEVs and SEVs were resuspended in

EV-depleted media at 1×10^6 /mL and added to BeWo cells for an additional 24 hours. At the time of harvest, cell media were collected, and cells released with 0.25% trypsin. Cells were washed and collected as pellets for total DNA measurement and RNA isolation and sequencing. Cell media was spun at 500G to clear cell debris and the supernatant was stored for future hormone measurement. Hormones were measured by Penn Fertility Care using Elecsys HCG+ β (Cat# 03271749, Roche Diagnostics) and Elecsys Progesterone III (Cat# 07092539, Roche Diagnostics).

Extravillous trophoblasts (EVTs) were isolated from fresh first trimester placenta based on an EVT outgrowth-based protocol established by Gram et al. (78-82). In brief, villous tissue was finely minced and cultured at 37°C and 5% CO₂ in RPMI 1640 media with 20% FBS. After attachment, EVT outgrowth occurs, and those cells were isolated. Isolated EVT cells were confirmed by staining for HLA-G and CK7. EVT invasion was measured using the MilliporeSigma Chemicon QCM Collagen Cell Invasion Assay (Cat# ECM558). EVT cells were added to the invasion plate with EV-depleted media and large and small EVs at 1×10^6 /mL. Cells were incubated at 37°C and 5% CO₂ for 48 hours. Cells that invaded into the second chamber were quantified using a fluorescent plate reader (SpectraMax).

EV mRNA sequencing

Total RNA was isolated from large and small EVs isolated from 500 μ L of plasma. Isolated EVs were treated with RNaseA (0.02 μ g/ μ L) (Invitrogen Cat# 12091021) for 20 minutes at 37°C to degrade extravesicular RNA. Enzyme activity was stopped by freezing samples at -80°C for 5 minutes and immediate resuspension in Trizol. Nucleic acids were isolated via BCP co-incubation, precipitated by isopropanol, and washed in ethanol. mRNA

libraries were prepared from total RNA using the SMART-Seq protocol (83). Briefly, RNA was reverse transcribed using Superscript II (Invitrogen, Cat#18064014). The cDNA was amplified with 20 cycles and cleaned up with AMPure XP beads (Beckman Coulter Cat#A63881). cDNA was quantified with Qubit dsDNA HS Assay Kit (Life Technologies, Inc. Cat#Q32851) and 2ng of each sample was used to construct a pool of uniquely indexed samples (Illumina Cat# FC-131-1096). A second amplification was performed with 12 cycles and cleaned up with AMPure XP beads. The final library was sequenced on a NextSeq 1000. Data were mapped against the hg19 genome using RSEM and normalized to transcripts per million (tpm)(84).

While there are differentially expressed transcripts compared to Controls, we chose not to focus on these because the statistical threshold (adjusted p-value <0.05) was rarely met due to high intrasubject variability. Instead, we only considered genes that are uniquely expressed in either Controls or COVID-19 cases. First, we filtered genes to those with greater than 5 transcripts per million and considered these expressed genes. Unique genes were determined to have no expression in any sample of the given group after filtration. Then only genes with expression in the majority (50% or more) of the samples in the comparison group were considered. As an example, a gene is uniquely expressed in R1 compared to Controls if all Controls samples have no expression and 5 or more samples (n=10) in R1 have expression (data in Table 5&6). Unique genes expressed by Controls but not a COVID-19 group are listed in Figure 6. These strict criteria identified genes in each group that are uniquely expressed. Those genes were considered for subsequent analysis.

EV mtDNA measurement

mtDNA was isolated and quantified from large and small EVs by TaqMan-based quantitative polymerase chain reaction (qPCR). We determined 2×10^7 LEV and 3×10^9 SEVs

were necessary to reliably and robustly measure mtDNA. The assay quantified mitochondrial-encoded human NADH: ubiquinone oxidoreductase core subunit 1 (ND1) as previously described (85). The qPCR reactions were performed in triplicates using a QuantStudio 5 Real-time PCR System (Thermo Fisher) using the following thermocycling conditions: 95 °C for 20 s followed by 40 cycles of 95 °C for 1 s, 63 °C for 20 s, and 60 °C for 20 s. Serial dilutions of pooled human placenta DNA quantified for copies of ND1 (copies/ μ L) by digital PCR (dPCR) were used as a standard curve. The amount of mtDNA per EV was determined by normalizing the resulting abundance by the number of EVs in starting material. A correlation between the gestational age at infection and the abundance of EV mtDNA was tested.

Statistical Analysis

GraphPad Prism was used to generate graphs and statistically test data. Subject demographics and outcomes were tested by a chi-squared test and considered statistically different if $p < 0.05$. Data was tested for normality and either parametric or non-parametric tests were used to determine significance. Data points were identified as outliers and removed if they exceeded two times the standard deviation from the mean. A one-way ANOVA tested for a difference within all groups and subsequent post-hoc t-tests or Kruskal-Wallis determined significance of each COVID-19 group compared to Controls. Pearson's correlation was used to determine a correlation between gestational age at infection and mtDNA abundance. A p-value less than 0.05 was considered significant. Differential gene expression was determined to be significant if the adjusted p-value was less than 0.05 and the fold change greater than 1.5.

List of Supplementary Materials

Supplemental Table 1. Placenta RNAseq DEG

Supplemental Table 2. EV Transcripts Associated with Adverse Pregnancy Outcomes

References and Notes

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Demographics	Controls (n=32)	R1 (n=15)	R2 (n=20)	R3 (n=22)	AI (n=21)
Maternal age (years)	21 – 42 (Mean: 31.8)	25 – 43 (Mean: 31.7)	18 – 39 (Mean: 29.4)	21 – 42 (Mean: 30.5)	20 – 40 (Mean: 29.4)
Race: White	50%	46.7%	34%	27.3%	9.5%*
Race: Black	41%	46.7%	55%	68.2%*	76.2%*
Race: Asian	9.4%	0%	10%	0%	9.5%
Race: Other/Unknown	0%	6.7%	0%	4.5%	4.8%
GA at delivery (weeks)	37.3 – 41.1 (Mean: 39.3)	35.1 – 40.3 (Mean: 39.1)	33.7 – 41.3 (Mean: 38.4)	32.3 – 39.7 (Mean: 37.5)*	31 – 41 (Mean: 38.1)*
COVID severity	Controls (n=32)	R1 (n=15)	R2 (n=20)	R3 (n=22)	AI (n=21)
Symptomatic	N/A	73.3%	85%	81.8%	14.3%
Asymptomatic	N/A	26.7%	15%	18.2%	85.7% ⁺
Hospitalized	N/A	13.3%	0%	31.8%	4.8%
Pregnancy outcomes	Controls (n=32)	R1 (n=15)	R2 (n=20)	R3 (n=22)	AI (n=21)
GHTN	18.8%	53.3%* ⁺	20%	13.6%	9.5%
CHTN	3.1%	0%	5%	13.6%	4.8%
Preeclampsia	0%	0%	5%	18.2%*	19%*
Spontaneous PTB	0%	0%	20%*	4.5%	4.8%
Medically indicated PTB	0%	6.7%	10%	18.2%*	19%*
IUGR	6.3%	13.3%	0%	9.1%	4.8%
IUFD	0%	0%	0%	4.5%	0%
Placenta Pathology	Control (n=26)	R1 (n=9)	R2 (n=17)	R3 (n=19)	AI (n=19)
MVM or FVM	26.9%	100%*	71%*	63%*	82.6%*
MVM	15%	44%*	47%*	37%	53%*
High grade MVM	0%	11%	18%*	16%*	11%
FVM	15%	78%*	59%*	42%*	47%*
>10% perivillous fibrin deposition	0%	11%	18%*	21%*	5%

Table 1. Subject demographics, pregnancy outcomes, and placenta pathology

Subjects enrolled in the COMET study formed five groups (controls, resolved infection in the 1st trimester (R1), 2nd trimester (R2), and 3rd trimester (R3), and active infection (AI). Maternal demographics including maternal age, race, and gestational age (GA) at birth are reported. The severity of COVID-19 during their pregnancy, incidence of pregnancy complications (gestational hypertension (gHTN), chronic hypertension (cHTN), preeclampsia, spontaneous and medically indicated preterm birth (PTB), intrauterine growth restriction (IUGR), and intrauterine demise (IUFD)) and placental pathology (maternal vascular malperfusion (MVM), fetal vascular malperfusion (FVM) and perivillous fibrin deposition) are reported * p<0.05 chi-squared test compared to Controls.

Control v. R1 Placenta Transcriptome				
canonical pathway	FC	Z-score	p-value	# genes
Tumor Microenvironment Pathway	4.61	1	0.00	4
Osteoarthritis Pathway	2.83	nc	0.00	3
Hematopoiesis from Multipotent Stem Cells	1.94	nc	0.01	1
IL-17 Signaling	1.86	nc	0.01	2
Granulocyte Adhesion and Diapedesis	1.85	nc	0.01	2
Hepatic Fibrosis / Hepatic Stellate Cell Activation	1.83	nc	0.01	2
Differential Regulation of Cytokine Production in Macrophages and T Helper Cells by IL-17A and IL-17F	1.76	nc	0.02	1
Agranulocyte Adhesion and Diapedesis	1.75	nc	0.02	2
Pyrimidine Deoxyribonucleotides De Novo Biosynthesis I	1.66	nc	0.02	1
Differential Regulation of Cytokine Production in Intestinal Epithelial Cells by IL-17A and IL-17F	1.66	nc	0.02	1
Control v. R3 Placenta Transcriptome				
canonical pathway	FC	Z-score	p-value	# genes
Hepatic Fibrosis / Hepatic Stellate Cell Activation	2.95	nc	0.00	4
Tumor Microenvironment Pathway	2.07	nc	0.01	3
ILK Signaling	1.93	nc	0.01	3
RAR Activation	1.91	nc	0.01	3
RHO GDI Signaling	1.85	nc	0.01	3
Autophagy	1.85	nc	0.01	3
VDR/RXR Activation	1.84	nc	0.01	2
Estrogen Receptor Signaling	1.81	-1	0.02	4
TR/RXR Activation	1.78	nc	0.02	2
AMPK Signaling	1.71	nc	0.02	3
Control v. AI Placenta Transcriptome				
canonical pathway	FC	Z-score	p-value	# genes
Phagosome Formation	16.90	-5.44	1.26E-17	17
CREB Signaling in Neurons	11.50	-3.78	3.16E-12	26
Hepatic Fibrosis / Hepatic Stellate Cell Activation	11.00	nc	1.00E-11	26
GP6 Signaling Pathway	11.00	-3.00	1.00E-11	37
Atherosclerosis Signaling	10.80	nc	1.58E-11	41
G-Protein Coupled Receptor Signaling	9.46	-3.79	3.47E-10	19
Breast Cancer Regulation by Stathmin1	8.88	-4.23	1.32E-09	19
Pulmonary Fibrosis Idiopathic Signaling Pathway	8.56	-2.60	2.75E-09	23
Role of Macrophages, Fibroblasts and Endothelial Cells in Rheumatoid Arthritis	8.50	nc	3.16E-09	42
Tumor Microenvironment Pathway	6.88	-2.00	1.32E-07	54

Table 2. Placenta RNAseq canonical pathways

Ingenuity pathway analysis was used to identify altered pathways in placenta collected from COVID-19 groups compared to control. The top 10 altered pathways are listed for each comparison and the log 2-fold change (FC), z-score, p-value, and number of pathway associated genes are reported. In the event a Z-score is not calculated, nc is reported.

BeWo Transcriptome Following Exposure to Control v. AI EVs			
gene symbol	gene name	FC	adj p-value
<i>RNU6-942P</i>	RNA, U6 Small Nuclear 942, Pseudogene	4.41	1.88E-02
<i>RN7SKP203</i>	RN7SK Pseudogene 203	2.12	4.11E-03
<i>FP236383.2</i>	unknown	1.93	7.35E-07
<i>AC068987.3</i>	unknown	1.77	1.46E-02
<i>HIST1H2AE</i>	H2A clustered histone 8	1.74	2.97E-04
<i>AC092127.2</i>	unknown	1.73	1.69E-04
<i>HIST2H2BF</i>	H2B clustered histone 18	1.70	2.97E-04
<i>HIST1H2BO</i>	H2B clustered histone 17	1.68	3.33E-06
<i>HIST1H2BN</i>	H2B clustered histone 15	1.65	3.01E-21
<i>HIST1H2AL</i>	H2A clustered histone 16	1.62	1.18E-02

Table 3. BeWo RNAseq DEG

Transcriptomics of BeWo cells following exposure to EVs identified differentially expressed genes when comparing exposure to AI vs. Controls EVs. The top 10 DEGs are listed for each comparison including the log 2-fold change (FC) and adjusted p-value for each comparison.

BeWo Transcriptome Following Exposure to Control v. AI EVs			
canonical pathway	FC	p-value	# genes
NAD Signaling Pathway	4.99	0.00	4
Granzyme A Signaling	3.86	0.00	2
DNA Methylation and Transcriptional Repression Signaling	3.19	0.00	2
Transcriptional Regulatory Network in Embryonic Stem Cells	2.95	0.00	2
NER (Nucleotide Excision Repair, Enhanced Pathway)	2.39	0.00	2
Ferroptosis Signaling Pathway	2.19	0.01	2
Sirtuin Signaling Pathway	1.53	0.03	2
nNOS Signaling in Neurons	1.37	0.04	1

Table 4. BeWo RNAseq pathway

Ingenuity pathway analysis identified canonical pathways altered following AI EV exposure compared to Controls EV exposure. The top 10 pathways are listed with the log 2-fold change (FC), p-value and number of pathway associated genes.

Uniquely Expressed Transcripts in COVID LEVs					
Gene Symbol	Gene Name	Group(s) with Unique Expression	Cellular Function	High Trophoblast Expression	Increased in Pregnancy
<i>YY1AP1</i>	YY1 Associated Protein 1	R1, R2	Transcription		
<i>MOSPD1</i>	Motile Sperm Domain Containing 1	R1, AI	Transcription		
<i>RYBP</i>	RING1 And YY1 Binding Protein	R1, AI	Transcription	Y	
<i>H1-4</i>	H1.4 linker histone, cluster member	R2, R3	Unknown		
<i>A2M</i>	Alpha-2-Macroglobulin	R1	Immune		
<i>KIFAP3</i>	Kinesin Associated Protein 3	R1	Chromosome Movement		Y(86)
<i>MXD1</i>	MAX Dimerization Protein 1	R1	Proliferation		
<i>PEX19</i>	Peroxisomal Biogenesis Factor 19	R1	Oxidative stress	Y	
<i>TOB1</i>	Transducer Of ERBB2, 1	R1	Proliferation		
<i>HK1</i>	Hexokinase 1	R2	Metabolism	Y	
<i>LY9</i>	Lymphocyte Antigen 9	R2	Immune		Y(87)
<i>POLR3C</i>	RNA Polymerase III Subunit C	R2	Nucleic acid binding activity		
<i>SPNS3</i>	SPNS Lysolipid Transporter 3, Sphingosine-1-Phosphate (Putative)	R2	Transporter Activity		
<i>WDR46</i>	WD Repeat Domain 46	R2	Nucleic acid binding activity		
<i>CDC34</i>	Cell Division Cycle 34, Ubiquitin Conjugating Enzyme	R3	Ubiquitination		
<i>FUNDC2</i>	FUN14 Domain Containing 2	R3	Metabolism		
<i>LINC-PINT</i>	Long Intergenic Non-Protein Coding RNA, P53 Induced Transcript	R3	Unknown		
<i>R3HCC1</i>	R3H Domain and Coiled-Coil Containing 1	R3	Nucleic acid binding activity		
<i>ATP5CKMT</i>	ATP Synthase C Subunit Lysine N-Methyltransferase	AI	Mitochondrial ATP synthesis		
<i>DTWD1</i>	DTW Domain Containing 1	AI	Translation		
<i>FBXL4</i>	F-Box and Leucine Rich Repeat Protein 4	AI	Ubiquitination		
<i>FRA10AC1</i>	FRA10A Associated CGG Repeat 1	AI	Transcription		
<i>FOXP1</i>	Forkhead Box P1	AI	Transcription		
<i>GOLGA4</i>	Golgi A4	AI	Protein and lipid transport	Y	Y(86)
<i>GPBP1L1</i>	GC-Rich Promoter Binding Protein 1 Like 1	AI	Transcription		
<i>IL1R2</i>	Interleukin 1 Receptor Type 2	AI	Immune		

<i>KIAA1143</i>	KIAA1143	AI	Unknown		
<i>LINC01410</i>	Long Intergenic Non-Protein Coding RNA 1410	AI	Unknown		
<i>NEAT1</i>	Nuclear Paraspeckle Assembly Transcript 1	AI	Transcription	Y	
<i>WDR26</i>	WD Repeat Domain 26	AI	Cell cycle progression and gene regulation	Y	
<i>XRNI</i>	5'-3' Exoribonuclease 1	AI	mRNA degradation		Y(88)
<i>ZNF638</i>	Zinc Finger Protein 638	AI	Transcription		

Table 5. Unique genes identified in LEVs isolated from COVID-19 cases

Transcripts are carried by LEVs isolated from COVID-19 cases that are absent in Controls. The listed transcripts are not detected in LEVs isolated from Controls but are present in most samples in the identified COVID-19 groups. The cellular function, expression in trophoblasts, and pregnancy associated expression of each transcript is listed as well (reference listed).

Uniquely Expressed Transcripts in COVID SEVs					
Gene Symbol	Gene Name	Group with Unique Expression	Cellular Function	High Trophoblast Expression	Increased in Pregnancy
<i>MYL4</i>	Myosin Light Chain 4	R1, AI	Motor Protein		Y(89)
<i>C18orf32</i>	Chromosome 18 Open Reading Frame 32	R1, AI	Immune	Y	
<i>CAPG</i>	Capping Actin Protein, Gelsolin Like	R2, AI	Motor Protein		
<i>CTSS</i>	Cathepsin S	R3, AI	Immune		Y(90)
<i>CYREN</i>	Cell Cycle Regulator Of NHEJ	R1	DNA Repair		
<i>CCDC124</i>	Coiled-Coil Domain Containing 124	R1	Transcription		
<i>CD27-ASI</i>	CD27 Antisense RNA 1	R1	Immune		
<i>GNB2</i>	G Protein Subunit Beta 2	R1	G Protein Signaling		
<i>PHF5A</i>	PHD Finger Protein 5A	R1	Immune	Y	
<i>RBM8A</i>	RNA Binding Motif Protein 8A	R1	Transcription	Y	
<i>SIGMAR1</i>	Sigma Non-Opioid Intracellular Receptor 1	R1	Calcium Signaling	Y	
<i>CSF3R</i>	Colony Stimulating Factor 3 Receptor	R2	Immune	Y	
<i>ESD</i>	Esterase D	R2	Metabolism		
<i>GTF2IRD2</i>	GTF2I Repeat Domain Containing 2	R2	Transcription		
<i>SLC1A5</i>	Solute Carrier Family 1 Member 5	R2	Metabolism	Y	Y(91)
<i>AUP1</i>	AUP1 Lipid Droplet Regulating VLDL Assembly Factor	R3	Ubiquitination	Y	
<i>MFF</i>	Mitochondrial Fission Factor	R3	Mitochondrial and Peroxisomal Fission		
<i>ARCNI</i>	Archain 1	AI	Vesicle	Y	
<i>CKAP2</i>	Cytoskeleton Associated Protein 2	AI	Proliferation		Y(92)
<i>CSNK2B</i>	Casein Kinase 2 Beta	AI	Metabolism		
<i>CYP27C1</i>	Cytochrome P450 Family 27 Subfamily C Member 1	AI	Metabolism		
<i>GRHL1</i>	Grainyhead Like Transcription Factor 1	AI	Epithelial Development	Y	
<i>HSP90B1</i>	Heat Shock Protein 90 Beta Family Member 1	AI	Molecular Chaperone	Y	Y(93)
<i>KIF5B</i>	Kinesin Family Member 5B	AI	Protein Binding Activity		
<i>LINC01123</i>	Long Intergenic Non-Protein Coding RNA 1123	AI	Unknown		
<i>LOC728323</i>	unknown	AI	Unknown		
<i>MANBAL</i>	Mannosidase Beta Like	AI	Membrane Protein		Y(94)

<i>MUC22</i>	Mucin 22	AI	Membrane Protein		
<i>MYLK</i>	Myosin Light Chain Kinase	AI	Contractile Activity		
<i>OR4F17</i>	Olfactory Receptor Family 4 Subfamily F Member 17	AI	Vesicle		
<i>PARP9</i>	Poly (ADP-Ribose) Polymerase Family Member 9	AI	Immune		
<i>RNF2</i>	Ring Finger Protein 2	AI	Transcription		
<i>SAR1A</i>	Secretion Associated Ras Related GTPase 1A	AI	Vesicle	Y	
<i>SHC4</i>	SHC Adaptor Protein 4	AI	Proliferation		
<i>SRSF8</i>	Serine And Arginine Rich Splicing Factor 8	AI	Transcription	Y	
<i>TMCC2</i>	Transmembrane And Coiled-Coil Domain Family 2	AI	Metabolism	Y	
<i>TRIM4</i>	Tripartite Motif Containing 4	AI	Immune		
<i>TWF2</i>	Twinfilin Actin Binding Protein 2	AI	Actin and ATP Binding Site		
<i>VTAI1</i>	Vesicle Trafficking 1	AI	Vesicle		
<i>ZNF484</i>	Zinc Finger Protein 484	AI	Transcription		

Table 6. Unique genes identified in SEVs isolated from COVID-19 cases
SEVs isolated from COVID-19 cases carry transcripts that are absent in Controls. The listed transcripts are not detected in SEVs isolated from Controls but are present in most samples in the identified COVID-19 groups. The cellular function, expression in trophoblasts, and pregnancy associated expression of each transcript is listed as well.

Control v. R1						
Upstream Regulator of Placenta Transcriptome					EV Expression	
gene symbol	gene name	activation score	p-value	# targets	LEV	SEV
<i>IGF2</i>	Insulin Like Growth Factor 2	1.96	2.41E-05	4	N	Y
<i>JUN</i>	Jun proto-oncogene	nc	4.74E-07	7	Y*	Y
<i>UCP3</i>	Uncoupling Protein 3	nc	9.82E-06	2	N	N
<i>FOS</i>	Fos proto-oncogene	nc	2.01E-05	6	Y	Y
<i>P2RY1</i>	Purinergic Receptor P2Y1	nc	2.06E-05	2	N	N
<i>PRLH</i>	Prolactin Releasing Hormone	nc	2.06E-05	2	N	N
<i>FGF3</i>	Fibroblast Growth Factor 3	nc	3.52E-05	2	N	N
<i>LEPR</i>	Leptin Receptor	nc	4.13E-05	4	Y	Y
<i>GPX8</i>	Glutathione Peroxidase 8 (Putative)	nc	4.40E-05	2	N	N
<i>ELN</i>	Elastin	nc	6.45E-05	2	N	N
<i>LGALS1</i>	Galectin 1	nc	8.82E-05	3	Y	Y
<i>FGF19</i>	Fibroblast Growth Factor 19	nc	9.77E-05	3	N	N
<i>PER2</i>	Period Circadian Regulator 2	nc	1.02E-04	2	N	N
<i>USP38</i>	Ubiquitin Specific Peptidase 38	nc	1.02E-04	2	N	N
<i>FOXC2</i>	Forkhead Box C2	nc	1.04E-04	3	N	N
<i>PRKN</i>	Parkin RBR E3 Ubiquitin Protein Ligase	nc	1.11E-04	3	N	Y
<i>RARG</i>	Retinoic Acid Receptor Gamma	nc	1.15E-04	3	N	N
<i>GAL</i>	Galanin And GMAP Prepropeptide	nc	1.33E-04	2	N	N
<i>CD36</i>	CD36 Molecule	nc	1.60E-04	3	Y	Y
<i>PER1</i>	Period Circadian Regulator 1	nc	2.04E-04	2	N	N
Control v. R3						
Upstream Regulator of Placenta Transcriptome					EV Expression	
gene symbol	gene name	activation score	p-value	# targets	LEV	SEV
<i>TNF</i>	Tumor Necrosis Factor	1.34	2.49E-05	16	N	N
<i>CYP19A1</i>	Cytochrome P450 Family 19 Subfamily A Member 1	0.882	1.52E-04	4	N	N
<i>IL1B</i>	Interleukin 1 Beta	-0.02	1.27E-04	11	Y ⁺	Y
<i>ESR1</i>	Estrogen Receptor 1	-0.27	4.36E-05	13	N	Y
<i>PGR</i>	Progesterone Receptor	-1.12	1.08E-05	7	N	N
<i>AKT1</i>	AKT Serine/Threonine Kinase 1	-1.42	1.03E-05	7	N	Y
<i>AGT</i>	Angiotensin	-1.44	9.92E-05	10	N	N
<i>PPARG</i>	Peroxisome Proliferator Activated Receptor Gamma	-1.63	4.27E-05	8	N	N
<i>HIF1A</i>	Hypoxia Inducible Factor 1 Subunit Alpha	-1.81	7.57E-05	8	Y	Y
<i>TP53</i>	Tumor Protein P53	-1.9	8.31E-05	15	N	Y
<i>FASLG</i>	Fas Ligand	nc	7.57E-06	4	N	N
<i>SPZ1</i>	Spermatogenic Leucine Zipper 1	nc	1.94E-05	3	N	Y
<i>CXCR4</i>	C-X-C Motif Chemokine Receptor 4	nc	2.16E-05	4	Y	N
<i>PLD1</i>	Phospholipase D1	nc	5.25E-05	3	N	N
<i>IGF2</i>	Insulin Like Growth Factor 2	nc	6.81E-05	5	Y	Y
<i>ANGPTL6</i>	Angiopoietin Like 6	nc	9.25E-05	2	N	N
<i>MAP2K1</i>	Mitogen-Activated Protein Kinase Kinase 1	nc	1.36E-04	5	N	N
<i>PCBP1</i>	Poly(RC) Binding Protein 1	nc	1.72E-04	2	Y	Y
<i>IRS2</i>	Insulin Receptor Substrate 2	nc	1.85E-04	3	N	Y
<i>TRIM29</i>	Tripartite Motif Containing 29	nc	2.21E-04	2	N	N

Control v. AI						
Upstream Regulator of Placenta Transcriptome					EV Expression	
gene symbol	gene name	activation score	p-value	# targets	LEV	SEV
<i>FAS</i>	Fas Cell Surface Death Receptor	1.46	5.86E-12	33	Y	N
<i>GRN</i>	Granulin Precursor	1.39	3.73E-12	19	Y	Y
<i>EZH2</i>	enhancer of zeste 2 polycomb repressive complex 2 subunit	1.04	1.15E-12	36	N	N
<i>HRAS</i>	HRas Proto-Oncogene, GTPase	0.67	3.72E-11	45	N	N
<i>AHR</i>	aryl hydrocarbon receptor	0.53	1.16E-11	38	N	N
<i>CEBPB</i>	CCAAT enhancer binding protein beta	0.35	4.92E-13	45	N	N
<i>JUN</i>	Jun proto-oncogene	0.04	3.54E-12	39	N	Y
<i>STAT6</i>	Signal Transducer and Activator of Transcription 6	-0.07	6.20E-14	40	N	N
<i>IL13</i>	Interleukin 13	-0.91	6.95E-19	47	N	N
<i>TNF</i>	Tumor Necrosis Factor	-0.98	6.17E-21	110	N	N
<i>IL6</i>	Interleukin 6	-1.34	7.28E-16	58	N	N
<i>STAT3</i>	Signal Transducer and Activator of Transcription 3	-2.07	4.88E-13	50	N	Y
<i>IL10</i>	Interleukin 10	-2.15	2.74E-17	45	N	N
<i>IL1B</i>	Interleukin 1 Beta	-2.21	5.16E-16	71	Y	Y
<i>IL4</i>	Interleukin 4	-2.36	6.94E-20	81	N	N
<i>IGF1</i>	Insulin Like Growth Factor 1	-2.37	1.28E-11	42	N	Y
<i>MAFB</i>	MAF BZIP Transcription Factor B	-3.02	3.53E-11	16	N	N
<i>IFNG</i>	Interferon Gamma	-4.00	1.27E-16	86	N	Y
<i>TGFB1</i>	Transforming Growth Factor Beta 1	-4.56	2.01E-23	115	N	N
<i>AGT</i>	Angiotensinogen	-4.74	1.26E-18	67	N	Y

Table 7. Placenta RNAseq upstream regulators

Ingenuity pathway analysis predicted upstream regulators of transcriptional changes in COVID-19 placenta compared to control. The activation score, p-value and number of targets identified in the dataset are listed for the top 20 predicted biological upstream regulators. Expression of the same gene in large and small EVs is recorded. An asterisk identifies instances where the differential gene expression between Control and the COVID-19 group is statistically different ($p < 0.05$). A plus sign identifies a statistical difference of $p < 0.1$.

Control v. R1 Placenta Transcriptome			
gene symbol	gene name	FC	adj p-value
<i>GSI-600G8.3</i>	unknown	7.48	2.09E-04
<i>AC027288.3</i>	unknown	6.85	4.35E-04
<i>RORB</i>	RAR Related Orphan Receptor B	6.10	3.45E-05
<i>MMP12</i>	Matrix Metalloproteinase 12	5.20	2.22E-04
<i>OMD</i>	Osteomodulin	4.32	3.48E-03
<i>IGFBP6</i>	Insulin Like Growth Factor Binding Protein 6	2.40	7.81E-05
<i>UPK1B</i>	Uroplakin 1B	2.59	1.16E-05
<i>FNI</i>	Fibronectin 1	2.38	1.41E-06
<i>SDK1</i>	Sidekick Cell Adhesion Molecule 1	1.62	1.15E-04
<i>MTARC2</i>	Mitochondrial Amidoxime Reducing Component 2	-2.15	1.67E-05
Control v. R2 Placenta Transcriptome			
gene symbol	gene name	FC	adj p-value
<i>SCN9A</i>	Sodium Voltage-Gated Channel Alpha Subunit 9	-1.70	0.047
<i>MAP1LC3C</i>	Microtubule Associated Protein 1 Light Chain 3 Gamma	-3.37	0.008
Control v. R3 Placenta Transcriptome			
gene symbol	gene name	FC	adj p-value
<i>AC079777.1</i>	Unknown	6.12	4.31E-03
<i>STAT4</i>	Signal Transducer and Activator of Transcription 4	3.04	1.14E-03
<i>ADAR1</i>	Adenosine Deaminase RNA Specific B1	-1.77	1.08E-03
<i>MICAL3</i>	Microtubule Associated Monooxygenase, Calponin and LIM Domain Containing 3	-1.97	8.89E-06
<i>SYNPO2</i>	Synaptopodin 2	-2.37	5.27E-03
<i>SIGLEC6</i>	Sialic Acid Binding Ig Like Lectin 6	-2.46	1.37E-04
<i>PRELP</i>	Proline And Arginine Rich End Leucine Rich Repeat Protein	-2.52	1.64E-03
<i>ENO2</i>	Enolase 2	-3.13	4.69E-06
<i>MEG9</i>	Maternally Expressed 9	-3.52	4.31E-03
<i>MAP1LC3C</i>	Microtubule Associated Protein 1 Light Chain 3 Gamma	-3.80	1.37E-04
Control v. AI Placenta Transcriptome			
gene symbol	gene name	FC	adj p-value
<i>PRL</i>	Prolactin	6.30	6.45E-08
<i>PENK</i>	Proenkephalin	5.10	1.84E-05
<i>TMPRSS3</i>	Transmembrane Serine Protease 3	3.83	3.94E-08
<i>GNLY</i>	Granulysin	3.39	2.08E-05
<i>LAMB3</i>	Laminin Subunit Beta 3	2.99	2.40E-06
<i>SLC22A3</i>	Solute Carrier Family 22 Member 3	-2.16	2.45E-05
<i>BMP5</i>	Bone Morphogenetic Protein 5	-2.56	2.42E-05
<i>CLIC2</i>	Chloride Intracellular Channel 2	-2.67	2.04E-05
<i>P2RX1</i>	Purinergic Receptor P2X 1	-3.42	1.23E-05
<i>PTGER2</i>	Prostaglandin E Receptor 2	-3.45	1.33E-05

Supplemental Table 1. Placenta RNAseq DEG

Differentially expressed genes (DEG) were identified by comparing COVID-19 groups to controls. The top 10 DEGs are reported with the log 2-fold change (FC) and adjusted p-value. When comparing R2 to control, only 2 genes surpassed the thresholds.

Uniquely Expressed Transcripts Associated with Adverse Pregnancy Outcomes					
Gene Symbol	Gene Name	EV Population	Uniquely Expressed in Group(s)	No Expression in Group(s)	Adverse Pregnancy Outcome(s)
<i>CENPT</i>	Centromere Protein T	LEV	Control	R1	FGR(95)
<i>IMP3</i>	IMP U3 Small Nucleolar Ribonucleoprotein 3	SEV	Control	R1	PE (96)
<i>RAP1GDS1</i>	Rap1 GTPase-GDP Dissociation Stimulator 1	LEV	Control	R2	FGR (97)
<i>LRRC8B</i>	Leucine Rich Repeat Containing 8 VRAC Subunit B	LEV	Control	R2	PTB (98)
<i>RNF138</i>	Ring Finger Protein 138	LEV	Control	R2	FGR (97)
<i>APPL1</i>	Adaptor Protein, Phosphotyrosine Interacting with PH Domain And Leucine Zipper 1	LEV	Control	R2	FGR (99)
<i>PAFAH1B2</i>	Platelet Activating Factor Acetylhydrolase 1b Catalytic Subunit 2	LEV	Control	R2	PE, gHTN, FGR (100, 101)
<i>CCDC18</i>	Coiled-Coil Domain Containing 18	LEV	Control	R2	PE (102)
<i>RNASE1</i>	Ribonuclease A Family Member 1, Pancreatic	SEV	Control	R2	PE (103)
<i>METTL14</i>	Methyltransferase 14, N6-Adenosine-Methyltransferase Subunit	LEV	Control	R3	PE (104, 105)
<i>LIN54</i>	Lin-54 DREAM MuvB Core Complex Component	LEV	Control	R3	FGR (106)
<i>FPR2</i>	Formyl Peptide Receptor 2	LEV	Control	R3	FGR (107)
<i>ALDH1A1</i>	Aldehyde Dehydrogenase 1 Family Member A1	LEV	Control	R3	FGR (108)
<i>SIPR4</i>	Sphingosine-1-Phosphate Receptor 4	LEV	Control	AI	PE (109)
<i>THAP12</i>	THAP Domain Containing 12	LEV	Control	AI	FGR (97)
<i>TNFSF13</i>	TNF Superfamily Member 13	LEV	Control	AI	PE (110)
<i>PSG2</i>	Pregnancy Specific Beta-1-Glycoprotein 2	LEV	Control	R2, AI	PE, PTB, FGR (111-113)
<i>APBA3</i>	Amyloid Beta Precursor Protein Binding Family A Member 3	LEV	Control	R2, R3, AI	PE (114)
<i>MXD1</i>	MAX Dimerization Protein 1	LEV	R1	Control	PTB (115) PE (116)

<i>RYBP</i>	RING1 And YY1 Binding Protein	LEV	R1	Control	PE (117-119)
<i>A2M</i>	Alpha-2-Macroglobulin	LEV	R1	Control	PE (120) PTB (121)
<i>HK1</i>	Hexokinase 1	LEV	R1	Control	PE (122)
<i>CAPG</i>	Capping Actin Protein, Gelsolin Like	SEV	R2	Control	PE (123) PTB (124)
<i>SLC1A5</i>	Solute Carrier Family 1 Member 5	SEV	R2	Control	PE (91)
<i>XRN1</i>	5'-3' Exoribonuclease 1	LEV	AI	Control	PE (125)
<i>WDR26</i>	WD Repeat Domain 26	LEV	AI	Control	FGR (97)
<i>GPBP1L1</i>	GC-Rich Promoter Binding Protein 1 Like 1	LEV	AI	Control	FGR (106)
<i>FOXP1</i>	Forkhead Box P1	LEV	AI	Control	PE (126)
<i>NEAT1</i>	Nuclear Paraspeckle Assembly Transcript 1	LEV	AI	Control	PE (127) FGR (128)
<i>KIF5B</i>	Kinesin Family Member 5B	SEV	AI	Control	PE (129)
<i>CSNK2B</i>	Casein Kinase 2 Beta	SEV	AI	Control	FGR (130)
<i>MYLK</i>	Myosin Light Chain Kinase	SEV	AI	Control	PE (116, 131)
<i>HSP90B1</i>	Heat Shock Protein 90 Beta Family Member 1	SEV	AI	Control	PE (132) PTB (133)
<i>MUC22</i>	Mucin 22	SEV	AI	Control	FGR (134)
<i>GRHL1</i>	Grainyhead Like Transcription Factor 1	SEV	AI	Control	FGR (135)
<i>SHC4</i>	SHC Adaptor Protein 4	SEV	AI	Control	PE (136)
<i>CTSS</i>	Cathepsin S	SEV	R3, AI	Control	FGR (137)

Supplemental Table 2. EV Transcripts Associated with Adverse Pregnancy Outcomes

Many transcripts detected in EVs are previously reported to be differentially expressed in adverse pregnancy complications. The gene, EV subpopulation (small [SEV] or large [LEV] EVs), unique expression, and identified literature are listed in the table.

Figure 1

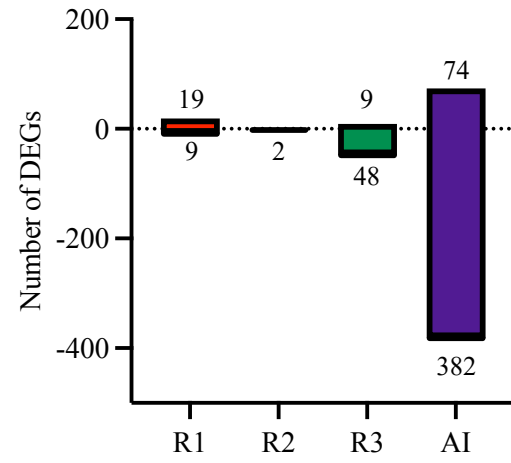


Figure 2

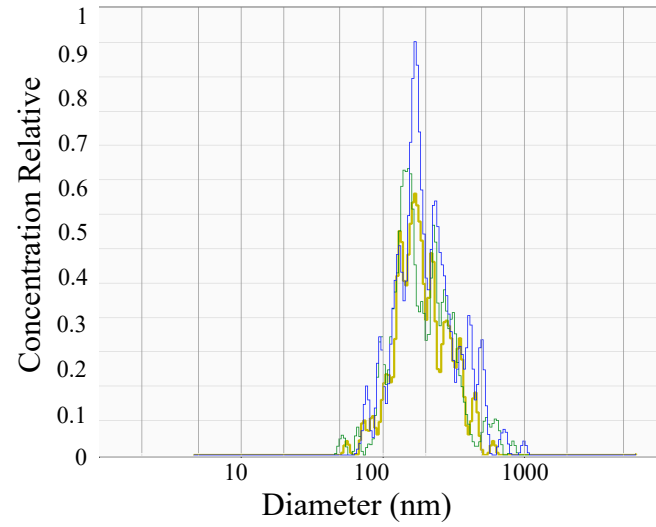
A Large EV



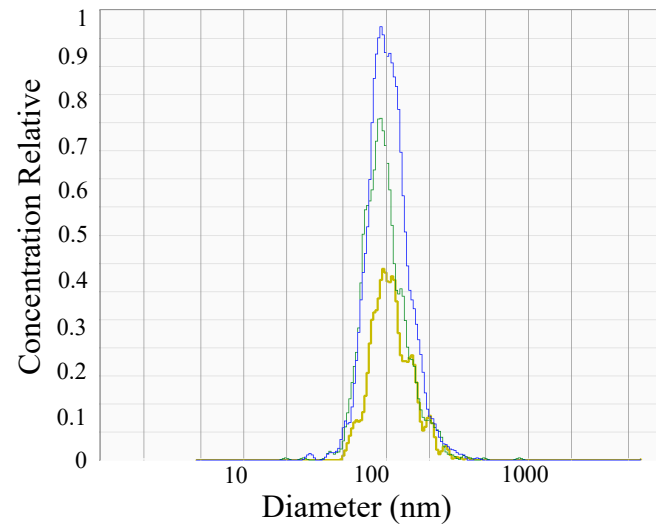
Small EV



B Large EVs



Small EVs



C

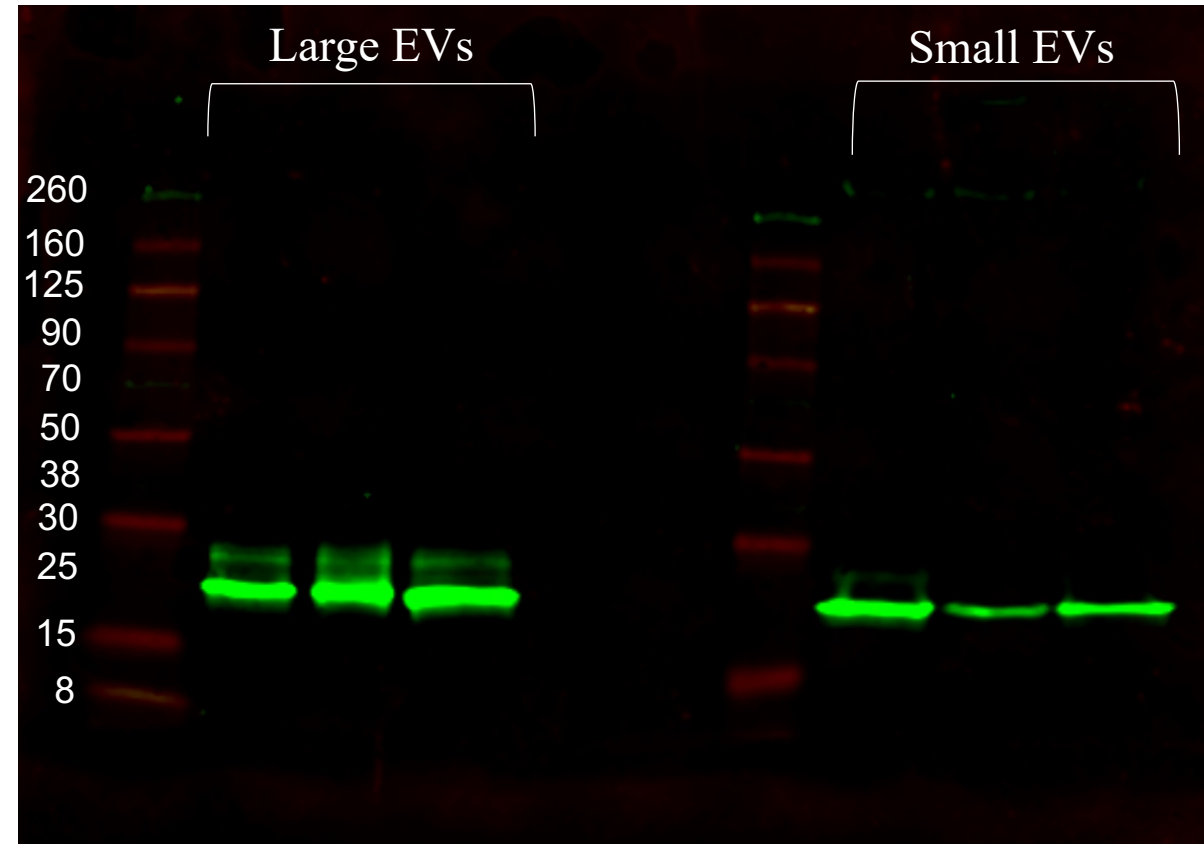
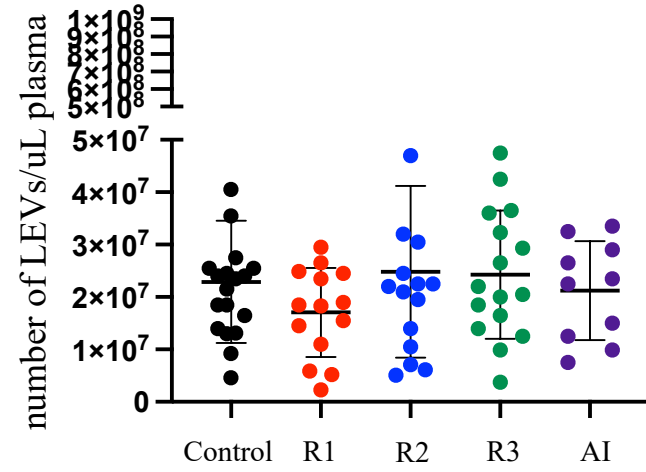
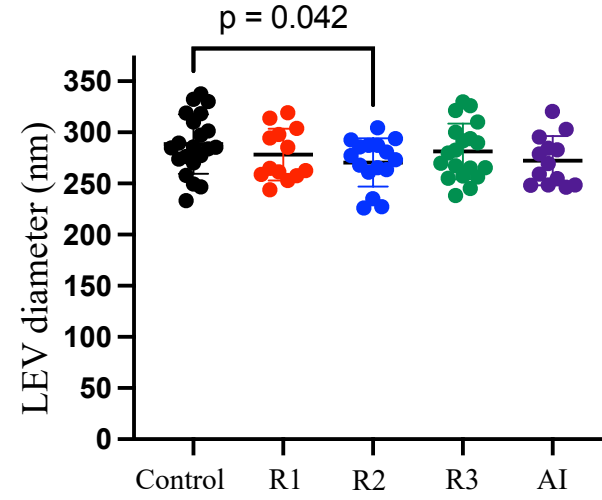


Figure 3

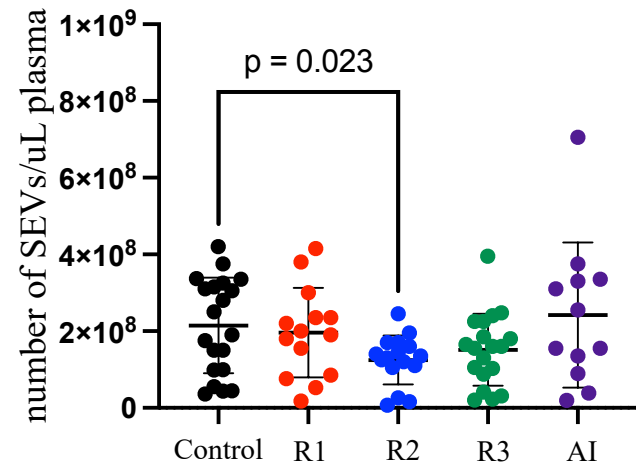
A



B



C



D

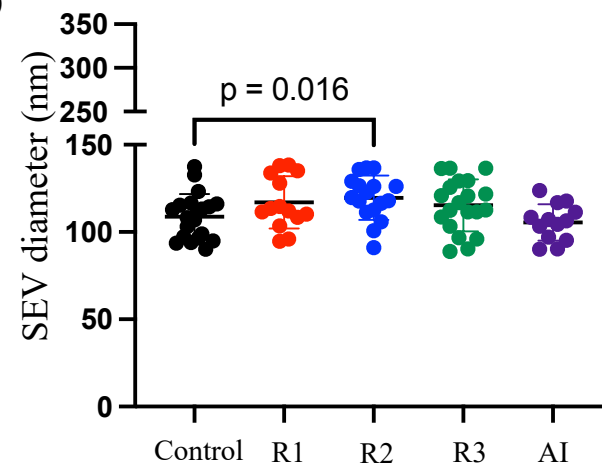
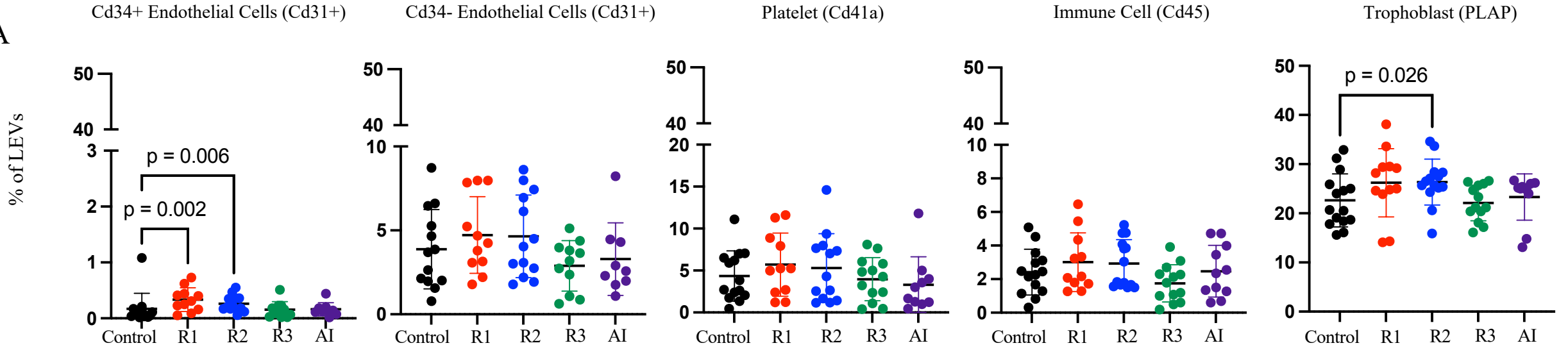


Figure 4

A



B

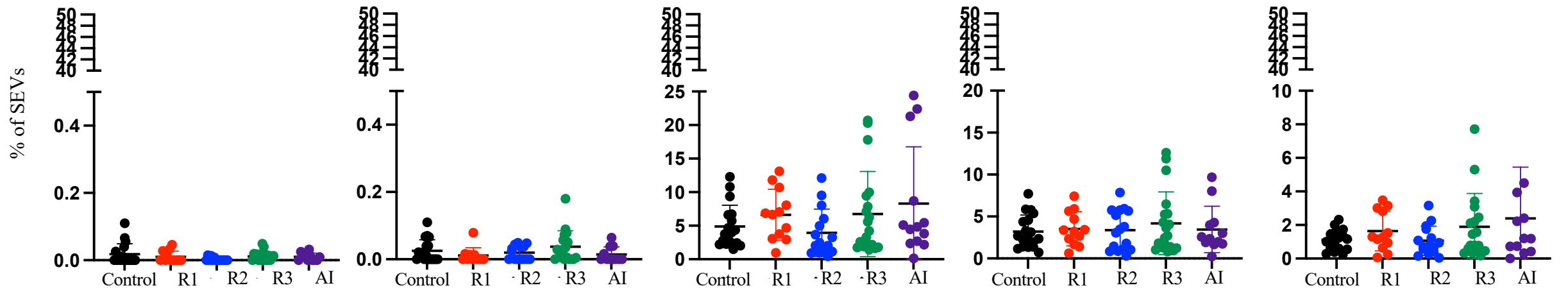


Figure 5

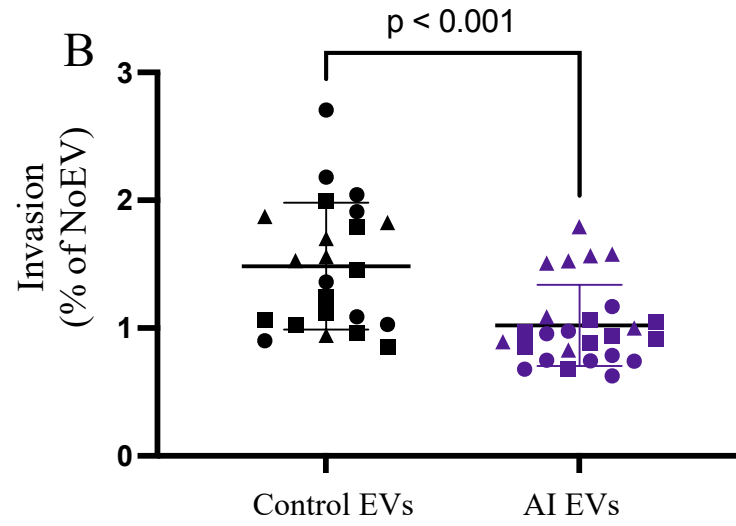
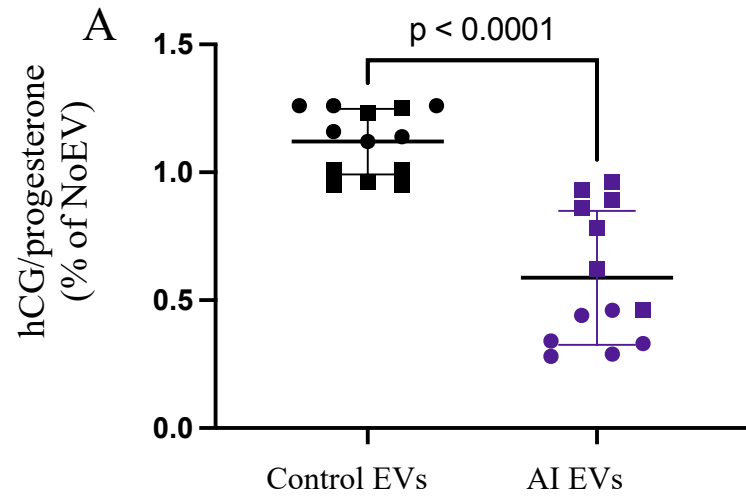


Figure 6

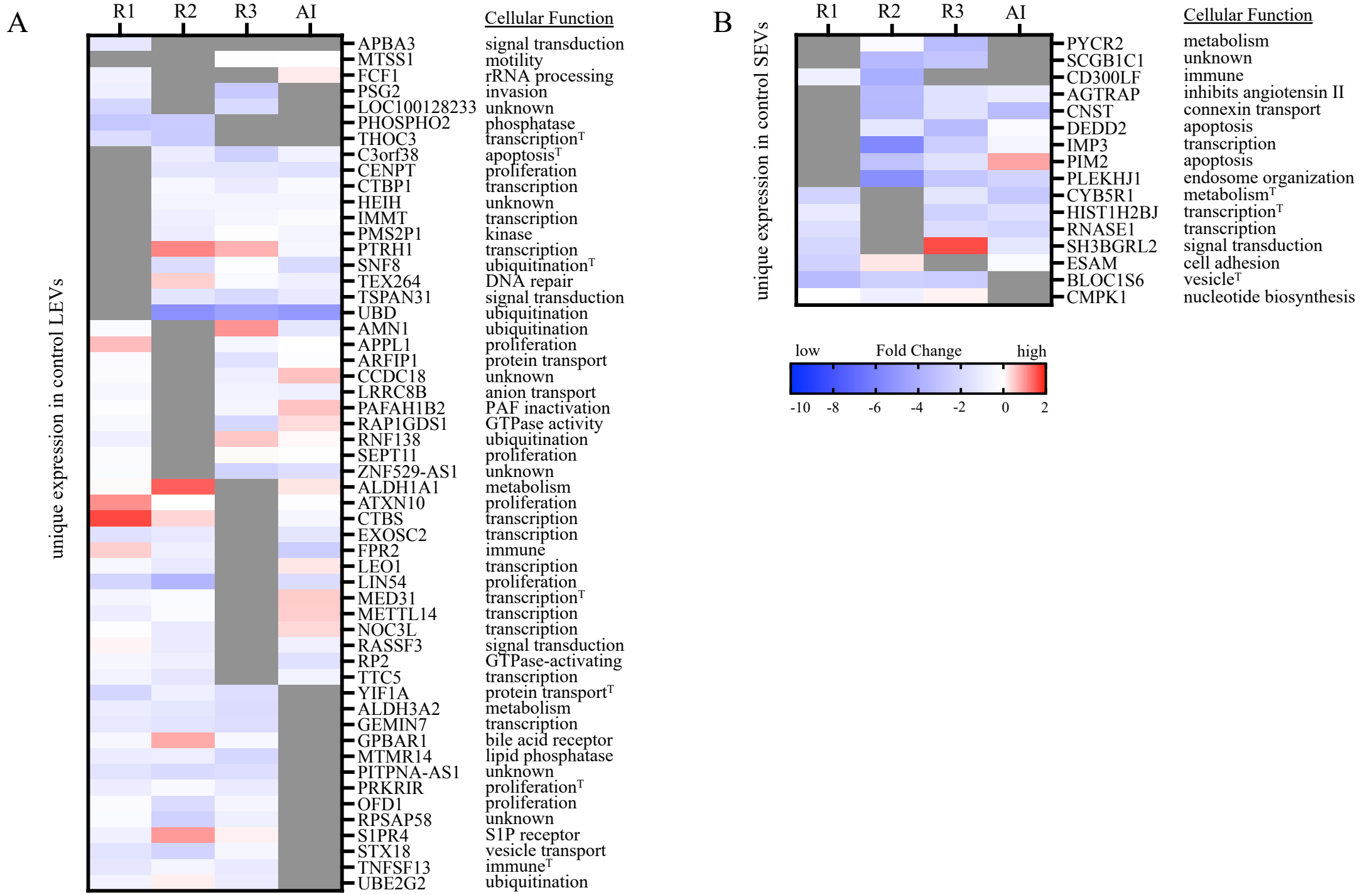


Figure 7

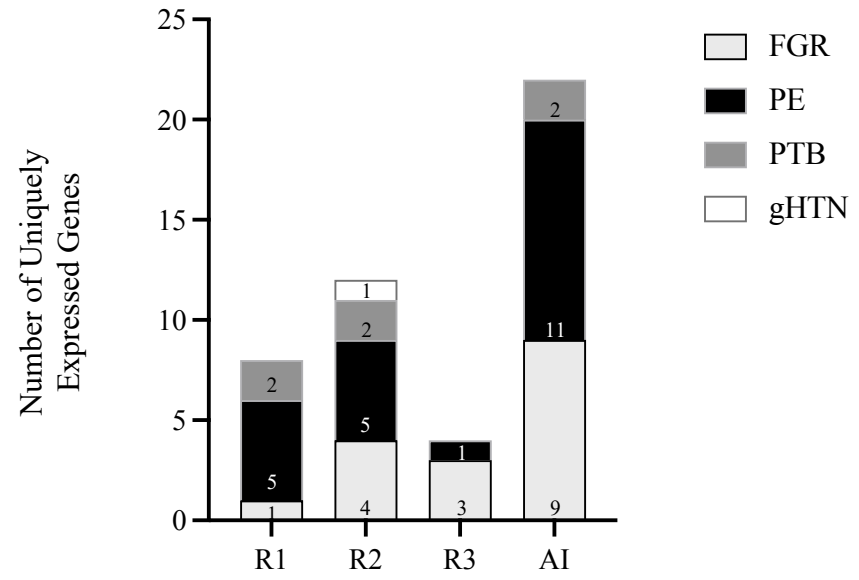


Figure 8

