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Oxygen and tissue culture affect placental gene expression

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	ACCEPTED MANUSCRIPT					
1	Oxygen and Tissue Culture affect Placental Gene Expression					
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### 10 Abstract

### 11

### 12 Introduction:

- 13 Placental explant culture is an important model for studying placental development and
- 14 functions. We investigated the differences in placental gene expression in response to tissue
- 15 culture, atmospheric and physiologic oxygen concentrations.

### 16 Methods

- 17 Placental explants were collected from normal term (38-39 weeks of gestation) placentae
- 18 with no previous uterine contractile activity. Placental transcriptomic expressions were
- 19 evaluated with GeneChip<sup>®</sup> Human Genome U133 Plus 2.0 arrays (Affymetrix).

### 20 Results

- 21 We uncovered sub-sets of genes that regulate response to stress, induction of apoptosis
- 22 programmed cell death, mis-regulation of cell growth, proliferation, cell morphogenesis,
- tissue viability, and protection from apoptosis in cultured placental explants. We also
- 24 identified a sub-set of genes with highly unstable pattern of expression after exposure to
- 25 tissue culture. Tissue culture irrespective of oxygen concentration induced dichotomous
- 26 increase in significant gene expression and increased enrichment of significant pathways
- and transcription factor targets (TFTs) including HIF1A. The effect was exacerbated by
- 28 culture at atmospheric oxygen concentration, where further up-regulation of TFTs including
- 29 PPARA, CEBPD, HOXA9 and down-regulated TFTs such as JUND/FOS suggest intrinsic
- 30 heightened key biological and metabolic mechanisms such as glucose use, lipid biosynthesis,
- 31 protein metabolism; apoptosis, inflammatory responses; and diminished trophoblast
- 32 proliferation, differentiation, invasion, regeneration, and viability.
- 33

## 34 Discussion

- 35 These findings demonstrate that gene expression patterns differ between pre-culture and
- 36 cultured explants, and the gene expression of explants cultured at atmospheric oxygen
- 37 concentration favours stressed, pro-inflammatory and increased apoptotic transcriptomic
- 38 response.
- 39

40

## 41 Keywords

- 42 Placenta; Gene Expression; atmospheric oxygen concentration; physiologic oxygen
- 43 concentration; Tissue Culture
- 44
- 45

### 46 **Abbreviations**

- 47 Atmospheric Oxygen Concentration: AOC; Physiologic Oxygen Concentration: POC;
- 48 AGE: Absolute Gene Expression; RGE: Relative Gene Expression; CHE: Consistent
- 49 High Expression; CLE: Consistent Low expression
- 50

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- 53 commercial, or not-for-profit sectors.

#### 1 Introduction:

2 In a recent publication [1] we reported on the damaging effect of atmospheric oxygen 3 concentration (AOC) on placental explant morphology and RNA quality. We showed 4 that while explants were viable after 6 days culture, there were more syncytial detachment and loss in explants cultured at the AOC (20%) than in physiological 5 6 oxygen concentration (POC) for term placentae (8% oxygen), and that the RNA quality and integrity of explants cultured at the AOC declined in tandem with 7 syncytiotrophoblast (STB) degeneration, damage and loss. This work confirmed 8 9 previous reports that while AOC has generally been used during culture, it could be 10 argued that 8% oxygen reflects in vivo physiology, and may provide optimal culture conditions for placental villous explants [2-7]. Yet, there has been no previous 11 systematic report on the effects of AOC or POC during explant culture on term placental 12 13 transcriptomic response to aid interpretations and discrimination between experimental treatment effect and culture oxygen effect. Our objectives therefore, were to investigate 14 the differences in placental transcriptomic changes in response to tissue culture and 15 oxygen, and the genetic alterations and pathways associated with POC and AOC 16 culture. 17

- 18
- 19 Methods
- 20 Tissue collection and culture

21 We collected placental explants from 6 placentae with no previous uterine contractile

22 activity following ethics permission (granted by the Hammersmith and Queen

23 Charlotte's & Chelsea Hospitals Research Ethics Committee) and written informed consent from patients. Details of methods used for placental collection, dissection, and 24 explant culture and viability assessment have been published elsewhere [1]. In brief, 25 approximately 2 cm<sup>3</sup> term (38-39 weeks of gestation) human placental explants were 26 randomly cut (3/placenta) immediately after delivery by elective Caesarean Section from 27 healthy looking areas about 5 cm away from the umbilical cord of normal pregnancies. 28 29 As previously described [1] micro explants (<50 mg wet weight) of villous tissue were 30 dissected from each sample (3 micro explants/sample) and cultured on 15mm diameter Netwell inserts with 74µm polyester mesh bottoms attached to polystyrene inserts 31 32 (Corning, UK) and incubated at the liquid-gas interface in POC (8% oxygen, 5% CO<sub>2</sub>) or AOC (95% air; 5% CO<sub>2</sub>). The micro explants were cultured in RPMI 1640 culture 33 medium (Invitrogen, UK) supplemented with 2 mM L-glutamine, 10% fetal bovine 34 35 serum, 100 IU/ml penicillin and 100 µg/ml streptomycin for 6 days. The culture media were replaced at days 2 and 4: the medium was placed in a sterile container and 36 37 exposed to the appropriate oxygen tension for 2 hours before the change of medium was done. The time for the change of medium was kept to the minimum needed (less 38 than 5 minutes). The explants at the end of the culture period were stored immediately 39 in RNAlater (Ambion) at -80°C. Pre-culture, 0 h con trol samples obtained from fresh 40 placentae were stored similarly in RNAlater within 30 minutes of delivery. 41

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#### 45 RNA and Microarray Preparation

46 RNA was extracted from explants cultured to time-point 120 h according to Chomczynski method [8] with TRIZOL reagent (Invitrogen, UK) and guality assessed. 47 Details of the methods for RNA extraction, quality and integrity assessments are 48 previously published [1]. Total RNA was processed into labelled cDNA with NuGEN™ 49 Ovation<sup>™</sup> RNA Amplification System V2 and FL-Ovation<sup>™</sup> cDNA Biotin Module V2 50 51 (Nugen). The resultant fragmented and labelled cDNA was added to the hybridisation cocktail in accordance with the NuGEN<sup>™</sup> guidelines for microarray hybridisation onto 52 Affymetrix GeneChip® Human Genome U133 Plus 2.0 arrays (sample per array) in 53 Affymetrix GeneChip® Hybridisation Oven 640 for 18 hours at 45°C. Features that 54 retained bound labelled cRNA after washing were visualized using the GeneChip® 55 Scanner 3000 (Affymetrix). The microarray data was published in the Gene Expression 56 57 Omnibus (GEO) repository with accession number GEO: GSE74446.

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#### 59 Array Quality Control and Processing

Quality control (QC) of the microarray raw data was assessed with Expression Console
(Affymetrix) for .CEL files integrity. Probes with unusual signal patterns or signal
strength and arrays showing low correlation between hybridization controls thus failed
this initial QC measures were excluded from further analysis. The .CEL raw data were
imported and processed with Robust Multi-array Average (RMA) into BRB-Array Tools
version 4.5.1 – Stable [9], and further processed using the R bionconductor packages
including Affy, annotate, annaffy, gcrma, globaltest, GO.db, lumi, ROC, simpleaffy,

67 bitops, car, gplots, GSA, impute, lars, matlab, pamr, randomForest. The arrays were log2-transformed and quantile-normalised to fit into linear model and a common scale to 68 generate expression measure for each probe set on each array. The spot filter analysis 69 70 was performed to remove spots whose signals were wrong due to small quantity of cDNA in the array, or errors during the scanning process. Furthermore, genes showing 71 minimal variation across the set of arrays were excluded from the analysis. Genes 72 73 whose expression differed by at least 1.5 fold from the median in at least 20% of the 74 arrays were retained.

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#### 76 Expression pattern analysis

We performed Relative Gene Expression (RGE) analysis for differentially expressed 77 genes between the AOC (experimental) and POC (control) classes using a random-78 variance t-test. The random-variance t-test permitted sharing of information among 79 genes about within-class variation without assuming that all genes have the same 80 81 variance [10]. This was supported with a Goeman's global test of whether the expression profiles differed between the classes by multiple permutation of the labels of 82 83 which arrays corresponded to which classes [11]. We also developed models to identify genes whose expression profiles could predict expression from explants cultured in 84 AOC at the p< 0.01 as assessed by the random variance t-test. Compound Covariate 85 Predictor (CCP), Diagonal Linear Discriminant Analysis (DLDA), Nearest Neighbor 86 Classification (NNC), and Support Vector Machines (SVM) with linear kernel [12-15], 87 were used to develop the models. Leave-one-out cross-validation (LOOCV) method 88 was used to compute mis-classification rate. The class labels were randomly permuted 89

90 (100 random permutations) and the entire LOOCV process was repeated. We further performed Absolute Gene Expression (AGE) analysis [16] using RankProd statistics 91 implemented in MEV\_4.9 [17], at FDR < 0.001 (Confidence (1-alpha): 99.9%) to identify 92 within class significant genes that were consistently expressed at high levels or low 93 levels respectively for pre-culture, AOC and POC classes. Genes that were consistently 94 expressed at high (CHE) levels in all the respective phenotype samples were classified 95 as positive significant genes. Alternatively, genes that were consistently expressed at 96 97 low (CLE) levels in all the respective phenotype samples were classified as negative significant genes. Genes that were not expressed consistently as either high or low in all 98 99 the samples were classified as non-significant genes.

100

### 101 Biological Significance analysis

102 Gene Ontology (GO), The National Cancer Institute, (Bethesda MD), experimentally verified transcription factor target (TFT) database and Kyoto Encyclopedia of Genes 103 and Genomes (KEGG) and BioCarta biological pathways were evaluated with functional 104 105 class scoring analysis as described by Pavlidis [18] to identify differential expression of 106 biologically relevant gene-sets between AOC and POC cultured explants. Significant gene-sets for differential expression were summarised with: (i) the Fisher (LS) statistics 107 (provides average log p values for the genes in the target class), (ii) the KS statistics 108 (Kolmogorov-Smirnov statistic computed on the p values for the genes in the target 109 class) and (iii) Gene Set Analysis (GSA) using Maxmean statistics [19] in BRB-array 110 Tools. For comparative analysis of significant gene-set enrichment within pre-culture 111

time zero, POC cultured and AOC explants, the respective gene pools from the AGE
analysis were tested with Benjamini-Hochberg statistics [20] implemented in
WebGestalt 2013 [21] to identify biologically relevant GOs, TFTs and KEGG biological
pathways.

116

117 **Results** 

### 118 Absolute Gene Expression in Pre-culture and Cultured Explants

As a preface to this study, we extensively evaluated RNA extraction and quality control 119 (QC) methods to ensure a high standard of guality for the RNA samples used (results 120 published in [1]). The biological replicate samples were representative of a realistic 121 122 application of microarrays in placental biology. Thus, following the normalisation and 123 quality control steps, 20,233 genes were used in further analyses. We performed AGE analysis using One Class RankProd statistics to identify 'consistently high expressed' 124 (CHE) and 'consistently low expressed' (CLE) significant genes respectively in Time 125 Zero (pre-culture) explants (designated as T0), and explants cultured in POC (8%) and 126 AOC (20% oxygen for 6 days. FDR Confidence (1-alpha) was set at 99.9% (FDR 127 128 <0.001).

Total of 635, 1207 and 1760 significant genes were consistently expressed in preculture (T0); POC and AOC cultured explants respectively (Suppl. Table 1). Of these,
224 genes were exclusively expressed in pre-culture samples (Figure 1A, sub-set 2). In
contrast, 69 genes (1 CHE and 68 CLE) were exclusively expressed in POC and 574

133	genes (292 CHE, and 282 CLE) in AOC oxygen cultured explants only (Figures 1 A & B;
134	Suppl. Table 1). We also identified 48 genes that were common to both pre-culture and
135	AOC explants only. Interestingly, CHE genes appeared insensitive to POC, but rather to
136	tissue culture and AOC. All CHE genes (except SIGLEC6) in POC explants were also
137	expressed consistently at high levels in AOC explants (Suppl. Table 2). There was a
138	core set of 770 consistently expressed significant genes (Figures 1 A and B sub-sets
139	3&5; Suppl. Tables 2 & 3) that were common to both AOC and POC samples only, and
140	these appear to be genes suggestively responding to tissue culture per se. No genes
141	were exclusively expressed consistently between pre-culture and POC samples only.

142

#### 143 Relative Gene Expression in Atmospheric and Physiological Oxygen

#### 144 **Concentration Cultured Explants**

145 We subsequently performed biological relevance network analysis [22] to determine the mutually biological relevance for performing relative gene expression analysis between 146 147 pre-culture, and explants cultured for 6 days. The results showed no biologically relevant mutual networks between pre-culture and the cultured explants (Suppl. Paper 148 1). We therefore performed the RGE analysis using a two-sample random-variance t-149 150 test for differentially expressed genes between the AOC (experimental) and POC (control) oxygen treated classes only and identified 157 significant genes (p<0.05). The 151 expression pattern was visualised with a Volcano plot (Figure 1C). We observed 88 up-152 153 regulated (Suppl. Table 4) and 69 down-regulated genes (Suppl. Table 5) in AOC 154 relative to POC explants. We further examined whether these genes could be

155	associated with explant culture in AOC. We thus performed class prediction modelling
156	incorporating Leave-one-out cross-validation and ROC curves. The analysis confirmed
157	the expression of 134 genes as significantly (p<0.01) associated with prolonged explant
158	culture at AOC (Suppl. Table 6). Of these, 12 genes were strongly associated (p<0.001)
159	with the AOC (Table 1). Three prediction algorithms: CCP, DLDA, and Bayesian
160	compound covariate predictor (BCCP) were used to generate a ROC curve (Figure 1D).
161	The analysis showed a very comparable ROC for all three algorithms with AUC of
162	0.82(CCP), 0.81(DLDA), 0.81(BCCP) (Figure 1D)
163	
164 165	Ontologies of Genes responding to Physiological and Atmospheric Oxygen Concentrations Differ

Table 2 shows a summary comparison of GOs, pathways and TFTs over-represented in 166 the pre-culture and cultured explants gene signatures (detailed in Suppl. Paper 2). 167 168 Figures 1 A & B sub-sets 3 and 5 contain a preserved set of genes that appears to 169 respond to tissue culture irrespective of oxygen concentration. We therefore examined 170 the GO, pathways and TFT associated with these preserved genes. The sub-set 3 genes (CHE genes present in both AOC and POC cultured explants only, irrespective of 171 172 oxygen concentration) significantly (p<0.001) enriched a cluster of cathartic ontologies 173 including programmed cell death, cell death, death, stress response, protein metabolic 174 process, electron transport activity, RNA translation and oxidoreductase activity (Table 175 3; Suppl. Table 10). The sub-set 3 genes also affected regulation of cytosol, 176 cytoplasmic and organelle parts of the placental cells. In contrast, sub-set 5 genes (CLE genes present in both AOC and POC explants only irrespective of oxygen 177

concentration) were mostly associated with the GO terms related to cellular
communication including multicellular organismal signalling, receptor binding, ionotropic
receptor, 3',5'-cyclic-GMP phosphodiesterase activity and ionotropic glutamate receptor
complex (Suppl. Table 11).

182

183 In addition, the AOC only CLE genes (Figure 1B sub-set 8) significantly enriched 184 specific GO terms including Passive transmembrane transporter activity, ion gated channel activity, calmodulin-dependent cyclic-nucleote phosphodiesterase activity, and 185 cation channel activity (Suppl. Table 12), while the CHE genes (Figure 1A sub-set 4) 186 187 further enriched proteolysis involved in cellular protein catabolic process, ubiquitindependent protein catabolic process, ligase activity, threonine-type peptase activity and 188 189 ubiquitin-protein ligase activity (Suppl. Table 13). There was no specific up-regulated 190 gene set for the POC explants. However, the POC only CLE genes (Figure 1B sub-set 191 7) significantly enriched specific molecular functions involved in multicellular organismal process and voltage-gated ion channel activity (Suppl. Table 14). 192

193

### 194 Effects of Atmospheric relative to Physiological Oxygen Concentrations

We further examined closely the effects of AOC relative to POC on placental
transcription factor target genes and biological pathway gene-sets. We used a twosample random variance T-test design, LS/KS permutation test and Efron-Tibshirani's
GSA maxmean to probe the National Cancer Institute, (Bethesda MD), experimentally
verified transcription factor target database. A total of 73 TFT gene-sets were

200	investigated and 15 (6 up-regulated and 9 down-regulated in AOC relative to POC)
201	were significantly (p<0.05) enriched (Table 4). The down-regulated TFT gene sets
202	including REL, ETV4, ATF3, STAT1, JUND and STAT5B are seemingly involved in cell
203	growth, proliferation, invasion, regeneration, differentiation, transformation, tissue
204	viability, protection from apoptosis, and glands development (Table 4). Conversely, the
205	up-regulated TFTs such as HIF1A, PPARA, CEBPD, STAT3, and CEBPE are mostly
206	associated with oxygen regulation, immune and inflammation responses; leptin
207	mediated response; lipid metabolism; suppression and mis-regulation of cell growth and
208	proliferation; cell morphogenesis; induction of apoptosis; and oxygen regulation (Table
209	4).
210	
211	Results from the RGE pathway analyses were consistent with the AGE GO and
212	transcription factor target gene sets analysis. For example, as expected Peroxisome,
212 213	transcription factor target gene sets analysis. For example, as expected Peroxisome, Nuclear Receptors in Lipid Metabolism and Toxicity, Glycerolipid metabolism, and
212 213 214	transcription factor target gene sets analysis. For example, as expected Peroxisome, Nuclear Receptors in Lipid Metabolism and Toxicity, Glycerolipid metabolism, and Reversal of Insulin Resistance by Leptin pathways were up-regulated (Tables 5; Suppl.
212 213 214 215	transcription factor target gene sets analysis. For example, as expected Peroxisome, Nuclear Receptors in Lipid Metabolism and Toxicity, Glycerolipid metabolism, and Reversal of Insulin Resistance by Leptin pathways were up-regulated (Tables 5; Suppl. Tables 15 and 16) in line with PPARA and STAT3 TFT gene-sets up-regulation.
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212 213 214 215 216 217	transcription factor target gene sets analysis. For example, as expected Peroxisome, Nuclear Receptors in Lipid Metabolism and Toxicity, Glycerolipid metabolism, and Reversal of Insulin Resistance by Leptin pathways were up-regulated (Tables 5; Suppl. Tables 15 and 16) in line with PPARA and STAT3 TFT gene-sets up-regulation. Similarly, RNA degradation, Nucleotide excision repair, and mRNA surveillance pathway were up-regulated in line with over-expression of CEBPE target gene-set.
212 213 214 215 216 217 218	transcription factor target gene sets analysis. For example, as expected Peroxisome, Nuclear Receptors in Lipid Metabolism and Toxicity, Glycerolipid metabolism, and Reversal of Insulin Resistance by Leptin pathways were up-regulated (Tables 5; Suppl. Tables 15 and 16) in line with PPARA and STAT3 TFT gene-sets up-regulation. Similarly, RNA degradation, Nucleotide excision repair, and mRNA surveillance pathway were up-regulated in line with over-expression of CEBPE target gene-set.
212 213 214 215 216 217 218 219	transcription factor target gene sets analysis. For example, as expected Peroxisome, Nuclear Receptors in Lipid Metabolism and Toxicity, Glycerolipid metabolism, and Reversal of Insulin Resistance by Leptin pathways were up-regulated (Tables 5; Suppl. Tables 15 and 16) in line with PPARA and STAT3 TFT gene-sets up-regulation. Similarly, RNA degradation, Nucleotide excision repair, and mRNA surveillance pathway were up-regulated in line with over-expression of CEBPE target gene-set.

221 Discussion

222 It is well accepted that placental explant culture is an important model for studying 223 placental transport, proliferation, differentiation, morphology, metabolism and endocrine functions under conditions akin to normal physiology. It is also clear that gestational age 224 mismatch with ex vivo oxygen concentration affects these placental functional 225 properties [23,24]. In this report, we have provided comprehensive evidence on 226 227 differences in transcriptomic expression underpinning general cell biology and 228 biochemical processes in response to placental explants culture at AOC and POC. A 229 key finding is the observation of the dichotomous increase in the number of significant 230 placental genes consistently expressed in line with oxygen concentration. Whereby, 231 more genes were consistently expressed at high levels than were expressed at low 232 levels. Interestingly, the disparate transcriptomic response not only affected parts or 233 extracellular environment of the placenta cells but also impacted on the elemental 234 activities of the gene products at the molecular level, such as binding or catalysis. 235

The uncovering of the placental gene sub-sets mediating programmed cell death; 236 response to stress; cell differentiation, and inhibition of cell proliferation was novel. 237 Albeit transcriptomic pathways at the moment, the findings are nonetheless exciting as 238 the current data provides evidence to encourage review of our understanding and 239 240 further study of the effects of culture conditions on trophoblast apoptosis, differentiation and proliferation. For, it is reported previously that trimester 1 villous explant culture in 241 242 approximately 3% oxygen stimulates increased trophoblast proliferation, while culture in 243 AOC appears to support proliferation but prevents invasiveness; and that AOC increases apoptosis in term villous explants [25-27]. 244

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Our current findings introgressively suggest that placental explant response to tissue 246 247 culture per se induces expression of genes with preponderance towards apoptosis irrespective of culture oxygen concentration. And that, the magnitude of the 248 249 programmed cell death could be sensitive to AOC. For example, we observed that AOC 250 was associated with down-regulation of trascription factor JUND target gene set 251 involved in apoptosis protection. Moreso, we observed relative up-regulation of PPARA 252 TFT gene set in explants cultured at AOC. Thus, considering that PPARA ligand activation has been linked to induction of apoptosis [28,29], it is plausible to surmise 253 254 that AOC could exacerbate placental programmed cell death through activation of the pro-apoptotic target gene sets via PPARA TFT. Furthermore, PPARA is associated with 255 lipid metabolism, cell differentiation, inhibition of cell proliferation, up-regulation of 256 257 immune and inflammation responses [26,27]. Therefore, it is equally consistent to extend the suggestion that poor trophoblast proliferation associated with AOC could be 258 mediated through activation of PPARA TFT gene set. 259

260

Certainly, HIF1A is well known to be constitutively expressed in the placenta to mediate hypoxic adaptation during placentation [30]. It has also been suggested that HIF-1A can be induced by factors other than hypoxia, including placental hormones, cytokines and growth factors, and well-oxygenated environment (~20% oxygen) [23,30,31]. We observed that HIF1A TFT gene sets were significantly enriched in response to AOC. The comparative analysis also showed that while no HIF1A target gene set was significantly enriched in pre-culture explants, the homolog V\$HIF1\_Q3 containing the

motif GNNKACGTGCGGNN target gene set was significantly enriched in both AOC and
POC samples. In addition, the target gene set for V\$HIF1\_Q5 containing the motif
CGTACGTGCNGB was also enriched in explants cultured in AOC. Thus, suggesting
that high oxygen could be an intrinsic regulatory mechanism for HIF activation in human
placenta.

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274 Our findings do not only provide evidence to confirm that well oxygenated (non-hypoxic) 275 environment regulates HIF1A target gene sets in the placenta, but also provide exciting 276 insights into the regulation of placental development through activator protein-1 (AP-1; 277 FOS/JUND). AP-1 proto-oncogenes have been linked to regulation of placental gene 278 expression in relation to oxygen concentration [32,33]. Our finding showed that JUND 279 and FOS target gene sets were also significantly enriched in response to AOC. While 280 our current data does not provide evidence on the interplay between HIF1A and AP1 pathways in the placenta, there are previous reports that suggest AP-1 may synergise 281 with HIF-1 to regulate hypoxic gene expression in the placenta [32-36]. It is therefore 282 plausible to suggest for further work that HIF1 and AP-1 target gene sets could provide 283 a co-regulatory response or feedback pathways in the regulation of placental 284 development. 285

286

Indeed, we have provided novel transcriptomic evidence not only to show that explant
culture per se could trigger placental gene sets that regulate programmed cell death
and stress response but also to support the consensus that AOC is pathological for
placental explant culture [6] by exacerbating a primed defective response.

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#### Tables

#### Table 1: Genes significantly associated with 20% oxygen culture

Symbol	Name	Fold-change	t-value	p-value
GNAZ	guanine nucleotide binding protein (G protein), alpha z	-1.95	-5.999	0.000103
	polypeptide			
NPPB	natriuretic peptide precursor B	-3.37	-5.917	0.000115
AIG1	androgen-induced 1	-1.46	-5.468	0.000221
C10orf90	chromosome 10 open reading frame 90	-1.95	-5.003	0.000444
TPH1	tryptophan hydroxylase 1	-1.69	-4.873	0.000543
PNMA2	paraneoplastic antigen MA2	-2.55	-4.855	0.000558
PRPS1	phosphoribosyl pyrophosphate synthetase 1	-1.85	-4.782	0.000626
MMP12	matrix metallopeptidase 12 (macrophage elastase)	-2.67	-4.546	0.00091
NEU1	sialidase 1 (lysosomal sialidase)	1.29	4.529	0.000936
C10orf12	chromosome 10 open reading frame 12	1.7	4.566	0.000882
LOC100129890	similar to hCG1750329	1.58	5.029	0.000427
ABHD4	abhydrolase domain containing 4	1.57	6.331	0.0000651

The prediction rule was defined by the inner sum of the weights  $(w_i)$  and log intensity expression  $(x_i)$  of significant genes.

398 399 Modelled predictors: A sample was classified to the class 20% oxygen if the sum was greater than the threshold; that is,

 $\sum_{i} w_i x_i$  > threshold. The threshold for the Compound Covariate predictor = -155.831; threshold for the Diagonal Linear Discriminant predictor = 318.343; threshold for the Support Vector Machine predictor = -7.845 (supplementary Table 4 contains full list of 20% oxygen concentration culture genes).

## Table 2: Enrichment analysis summary table

Enrichment	Biological	Molecular	Cellular	KEGG Pathways	Transcription Factor
	Processes	Process	Component		Targets
Con Total	40	11	40	42	251
8% Total	40	15	40	61	400
20% Total	40	23	40	121	497
Overlap					
Enriched in Pre-culture only	5	2	9	2	4
Enriched in 8% Oxygen only	2	1	4	0	7
Enriched in 20% Oxygen only	1	8	3	55	92
Enriched in all explants	32	8	30	35	235
Enriched in Pre-culture and 8% only	1	0	0	0	0
Enriched in Pre-culture and 20% only	2	1	1	5	12
Enriched in 8% and 20% only	5	6	6	26	158

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Details of enriched pathways and GO terms are provided in supplementary tables

Database	GO ID	Name	С	0	E	R	rawP	adjP
BP	GO:0032268	regulation of cellular protein	1250	64	27.85	2.3	1.89E-10	3.89E-07
		metabolic process						
BP	GO:0051246	regulation of protein metabolic process	1413	68	31.49	2.16	6.52E-10	4.47E-07
BP	GO:0006091	generation of precursor metabolites and energy	451	34	10.05	3.38	5.03E-10	4.47E-07
BP	GO:0006950	response to stress	2952	111	65.78	1.69	2.29E-09	1.18E-06
BP	GO:0016265	death	1706	74	38.01	1.95	9.60E-09	2.82E-06
BP	GO:0008219	cell death	1704	74	37.97	1.95	9.13E-09	2.82E-06
BP	GO:0009987	cellular process	12899	316	287.43	1.1	9.01E-09	2.82E-06
BP	GO:0022900	electron transport chain	148	17	3.3	5.15	3.28E-08	8.04E-06
BP	GO:0016071	mRNA metabolic process	613	37	13.66	2.71	3.52E-08	8.04E-06
BP	GO:0012501	programmed cell death	1545	67	34.43	1.95	5.90E-08	1.21E-05
MF	GO:0005515	protein binding	7337	215	155.05	1.39	8.23E-12	2.83E-09
MF	GO:0003723	RNA binding	854	41	18.05	2.27	7.36E-07	0.0001
MF	GO:0008092	cytoskeletal protein binding	638	32	13.48	2.37	5.46E-06	0.0006
MF	GO:0003743	translation initiation factor activity	50	8	1.06	7.57	9.04E-06	0.0006
MF	GO:0015078	hydrogen ion transmembrane transporter activity	101	11	2.13	5.15	9.26E-06	0.0006
MF	GO:0004129	cytochrome-c oxase activity	28	6	0.59	10.14	2.16E-05	0.0009
MF	GO:0016676	oxoreductase activity, acting on a	28	6	0.59	10.14	2.16E-05	0.0009
		heme group of donors, oxygen as acceptor						
MF	GO:0015002	heme-copper terminal oxase activity	28	6	0.59	10.14	2.16E-05	0.0009
MF	GO:0016675	oxoreductase activity, acting on a heme group of donors	29	6	0.61	9.79	2.68E-05	0.001
MF	GO:0005488	binding	11955	281	252.65	1.11	3.14E-05	0.0011
СС	GO:0005737	cytoplasm	9130	267	185.24	1.44	3.32E-20	8.80E-18
СС	GO:0044444	cytoplasmic part	6772	208	137.4	1.51	8.35E-15	1.11E-12
СС	GO:0005829	cytosol	2372	101	48.13	2.1	7.20E-14	6.36E-12
СС	GO:0044424	intracellular part	12237	304	248.28	1.22	2.98E-13	1.97E-11
СС	GO:0005622	intracellular	12564	306	254.92	1.2	7.36E-12	3.90E-10
СС	GO:0043226	organelle	10651	268	216.11	1.24	1.17E-09	5.17E-08
СС	GO:0043229	intracellular organelle	10636	267	215.8	1.24	2.00E-09	7.57E-08
СС	GO:0044446	intracellular organelle part	6725	189	136.45	1.39	5.69E-09	1.88E-07
СС	GO:0044422	organelle part	6812	190	138.21	1.37	9.62E-09	2.83E-07
СС	GO:0044464	cell part	14643	329	297.1	1.11	1.87E-08	4.58E-07

Table 3: Ontologies of Genes Re	esponding to Tissue Culture Irre	spective of Oxygen Concentration
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C: the number of reference genes in the category; O: the number of genes in the gene set and also in the category; E: the expected number in the category; R: ratio of enrichment; rawP: p value from hypergeometric test; adjP: p value adjusted by the multiple test adjustment

Transcription Factor	Function	Number	n-value	Maxmean Di
Gene-Sets	Tunction .	ofgenes	p-value	Waxinean Di
REL T00168	A proto-oncogene Involved in NF-	23	0.00101	(-)
REE_100100	κB transcription. Promotes B-	25	0.00101	()
	cell survival and proliferation and			
	lymphoma			
RELA_T00594	Involved in NF-kB dependant cellular	81	0.00417	(-)
	metabolism, chemotaxis. Modulates			
	immune responses. Positively associated			
	with cancer.			
SPI1_T02068	Activates gene expression during	84	0.00494	(-)
	myeloid and B-lymphoid cell			
	development. Regulates purine-rich			
	sequence and alternative splicing of			
	target genes.	0		
HOXA9_101709	Regulates gene expression, cell	9	0.005	(+)
EOG E00102	morphogenesis, cell differentiation	20	0.005	
FOS_100123	API transcription factor complex.	39	0.005	(-)
	differentiation, and transformation			
	Associated with apoptotic cell death			
HIF1A T01609	Associated with apoptotic cell death.	72	0.01	(+)
THTTA_101009		72	0.01	(+)
E1V4_100685	Activates matrix metalloproteinase		0.01188	(-)
	genes. Associated with invasion and			
CEBDE T0/883	Transcriptional mis regulation in cancer	7	0.01/83	(1)
CEDFE_104003	Transcriptional mis-regulation in cancer.		0.01463	(+)
PPARA_105221	Lipid metabolism, cell differentiation,	50	0.01561	(+)
	influence influence in the second sec			
	and initialinitation responses. Induces			
ATE3 T01313	Induced upon physiological stress in	11	0.02371	()
AII 5_101515	various tissues. A marker of	11	0.02571	(-)
	regeneration following injury.			
STAT1 T01492	Mediates and upregulates genes	48	0.0256	(-)
<u></u>	expression for cell viability. Induces		0.0200	
	cellular antiviral state.			
CEBPD_T00583	Growth suppression	21	0.02734	(+)
JUND T01978	AP1 transcription factor complex.	15	0.03	(-)
	Protects cells from p53-dependent			
	senescence and apoptosis.			
STAT5B_T05736	Mediates signal transduction from	23	0.03393	(-)
	cytokines and growth hormones.			
	Involved in TCR signalling, apoptosis,			
	mammary gland development.			
STAT3_T05694	Mediates responses to interleukins,	50	0.03948	(+)
	KITLG/SCF, LEP and other growth			
	factors. Cell cycle regulation.			

#### Table 4: Enriched Transcription Factor Target Gene Sets at high oxygen

Table shows 15 out of 73 investigated gene sets. LS/KS permutation test found 12 significant gene sets. Efron-Tibshirani's maxmean test found 6 significant gene sets (under 200 permutations). (+) and (-) represent respectively, up or down-regulated transcription factor target gene set in high oxygen explants (20%) relative to low oxygen (8%), as determined with Efron-Tibshirani's maxmean test (Detailed of gene-sets in suppl. Table)

Pathway	Pathway ID	Pathway description	Number	p-value	Maxmean
			of genes		
KEGG	hsa00030	Pentose phosphate pathway	28	0.00013	(+)
	hsa00561	Glycerolipid metabolism	51	0.00058	(+)
	hsa03420	Nucleotide excision repair	45	0.005	(+)
	hsa04710	Circadian rhythm	23	0.005	(+)
	hsa00310	Lysine degradation	44	0.00665	(+)
	hsa00400	Phenylalanine, tyrosine and tryptophan biosynthesis	5	0.00749	(+)
	hsa00620	Pyruvate metabolism	40	0.00913	(+)
	hsa00053	Ascorbate and aldarate metabolism	18	0.0111	(+)
	hsa04146	Peroxisome	81	0.015	(+)
	hsa03010	Ribosome	87	0.015	(+)
BioCarta	h_vitCBPathway	Vitamin C in the Brain	11	0.005	(+)
	h_ace2Pathway	Angiotensin-converting enzyme 2 regulates heart function	13	0.005	(+)
	h_leptinPathway	Reversal of Insulin Resistance by Leptin	11	0.005	(+)
	h_npp1Pathway	Regulators of Bone Mineralization	10	0.005	(+)
	h_plateletAppPathway	Platelet Amyloid Precursor Protein Pathway	14	0.01	(+)
	h_cardiacegfPathway	Role of EGF Receptor Transactivation by GPCRs in Cardiac Hypertrophy	18	0.01	(+)
	h_erkPathway	Erk1/Erk2 Mapk Signaling pathway	28	0.01354	(+)
	h_alkPathway	ALK in cardiac myocytes	37	0.01593	(+)
	h_nuclearRsPathway	Nuclear Receptors in Lipid Metabolism and Toxicity	35	0.03298	(+)
	h_akap95Pathway	AKAP95 role in mitosis and chromosome dynamics	12	0.035	(+)
KEGG	hsa05150	Staphylococcus aureus infection	53	0.00022	(-)
	hsa05323	Rheumatoid arthritis	88	0.00043	(-)
	hsa00590	Arachidonic acid metabolism	57	0.00379	(-)
	hsa04350	TGF-beta signaling pathway	84	0.00404	(-)
	hsa04610	Complement and coagulation cascades	71	0.005	(-)
	hsa00603	Glycosphingolipid biosynthesis - globo series	14	0.005	(-)
	hsa05014	Amyotrophic lateral sclerosis (ALS)	55	0.005	(-)
	hsa00232	Caffeine metabolism	8	0.00758	(-)
	hsa04940	Type I diabetes mellitus	44	0.01932	(-)
	hsa05144	Malaria	53	0.01969	(-)
BioCarta	h_compPathway	Complement Pathway	19	0.0001	(-)
	h_classicPathway	Classical Complement Pathway	14	0.0006	(-)
	h_antisensePathway	RNA polymerase III transcription	5	0.005	(-)
	h_mspPathway	Msp/Ron Receptor Signaling Pathway	7	0.00708	(-)
	h_eicosanoidPathway	Eicosanoid Metabolism	22	0.00967	(-)

## Table 5: Significant Pathway Gene Sets Associated with high Oxygen

h	_p38mapkPathway	p38 MAPK Signaling Pathway	36	0.01104	(-)
h	_lectinPathway	Lectin Induced Complement Pathway	12	0.01205	(-)
h	_il10Pathway	IL-10 Anti-inflammatory Signaling Pathway	13	0.01312	(-)
h	_alternativePathway	Alternative Complement Pathway	9	0.01335	(-)
h	_inflamPathway	Cytokines and Inflammatory Response	29	0.01494	(-)

Table contains the top 10 most significant up and down-regulated biological pathways from KEGG and BioCarta. Supplementary Tables 15 and 16 contain full list of the significantly enriched pathways associated with high oxygen. (+) and (-) represent respectively, up or down-regulated transcription factor target gene set determined with Efron-Tibshirani's maxmean test

#### Oxygen and Tissue Culture affect Placental Gene Expression

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### Figure 1: Expressed Genes in Explants Cultured at Physiologic and

### **Atmospheric Oxygen Concentrations**

Figure 1A shows overlap of consistent high level expressed (CHE) genes between pre-culture, and explants culture at AOC and POC. Figure 1B shows overlap between pre-culture CHE genes and cultured explants consistent low expressed (CLE) genes. Figure 1C shows a volcano plot of 157 significant genes (blue dots). Random variance model parameters at a = 1.31652, b = 17.03489, Kolmogorov-Smirnov statistic= 0.01 and a nominal significance level (dotted line) of each univariate test at p< 0.05 (210 exact permutations). Figure 1D shows the ROC curve from the Bayesian Compound Covariate Predictor for AOC associated genes. S1 – S8 = Gene Sub-sets 1 – 8 T0 = Pre-culture CHE genes; AOC = Atmospheric Oxygen Concentration; POC = Physiologic Oxygen Concentration; +ve = CHE; -ve = CLE.

NB: Print figure 1 in colour



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### Highlights: Oxygen and Tissue Culture affects Placental Gene Expression

- Gene expression patterns differ between pre-culture and cultured explants
- Tissue culture up-regulates apoptosis and response to stress genes in placenta
- Atmospheric Oxygen Concentration up-regulates HIF1A transcription target
   gene set
- Atmospheric Oxygen Concentration regulated genes favour apoptosis and inflammation

#### **Conflict of Interest Statement**

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.