Supporting information

RNA-Seq reveals conservation of function among the yolk sacs of human, mouse and chicken

Tereza Cindrova-Davies, Eric Jauniaux, Michael Elliot, Sungsam Gong, Graham J Burton, D. Stephen Charnock-Jones

Methods

Proteomics. 1D gel lanes were cut into 8 bands and each band was transferred into a 96 well PCR plate. The bands were cut into 1mm² pieces, destained, reduced (DTT) and alkylated (iodoacetamide) and subjected to enzymatic digestion with trypsin overnight at 37°C. After digestion, the supernatant was pipetted into a sample vial and loaded onto an autosampler for automated LC-MS/MS analysis.

All LC-MS/MS experiments were performed using a Dionex Ultimate 3000 RSLC nanoUPLC (Thermo Fisher Scientific Inc, Waltham, MA, USA) system and a QExactive Orbitrap mass spectrometer (Thermo Fisher Scientific Inc, Waltham, MA, USA). Separation of peptides was performed by reverse-phase chromatography at a flow rate of 300 nL/min and a Thermo Scientific reverse-phase nano Easy-spray column (Thermo Scientific PepMap C18, 2µm particle size, 100Å pore size, 75µm i.d. x 50cm length). Peptides were loaded onto a precolumn (Thermo Scientific PepMap 100 C18, 5µm particle size, 100Å pore size, 300µm i.d. x 5mm length) from the Ultimate 3000 autosampler with 0.1% formic acid for 3 minutes at a flow rate of 10 µL/min. After this period, the column valve was switched to allow elution of peptides from the pre-column onto the analytical column. Solvent A was water + 0.1% formic acid and solvent B was 80% acetonitrile, 20% water + 0.1% formic acid. The linear gradient employed was 2-40% B in 30 minutes.

The LC eluant was sprayed into the mass spectrometer by means of an Easy-Spray source (Thermo Fisher Scientific Inc.). All *m/z* values of eluting ions were measured in an Orbitrap mass analyzer, set at a resolution of 70000 and was scanned between *m/z* 380-1500. Data dependent scans (Top 20) were employed to automatically isolate and generate fragment ions by higher energy collisional dissociation (HCD, NCE:25%) in the HCD collision cell and measurement of the resulting fragment ions was performed in the Orbitrap analyser, set at a resolution of 17500. Singly charged ions and ions with unassigned charge states were excluded from being selected for MS/MS and a dynamic exclusion window of 20 seconds was employed.

Post-run, the data was processed using Protein Discoverer (version 1.4., ThermoFisher). Briefly, all MS/MS data were converted to mgf files and the files were then submitted to the Mascot search algorithm (Matrix Science, London UK) and searched against the UniProt human database (153168 sequences; 54677058 residues). Variable modifications of oxidation (M) and deamidation (NQ) and a fixed modification of carbamidomethyl (C) were applied and the peptide and fragment mass tolerances were set to 5ppm and 0.1 Da, respectively. A significance threshold value of $p<0.05$ and a peptide cut-off score of 20 were also applied. All the Mascot data from each lane was merged to give a single output file. Finally, Mascot .dat files were entered into the Scaffold software (Proteome Software, Oregon USA) so that differences between the two protein samples could be compared.

Two samples had very few recognizable peptide hits these samples were excluded from the subsequent analysis. Unique hits to identifiable proteins were counted and those with the same gene name in the identifier filed combined. The corresponding gene names for the proteins detected in 4 out of the 5 samples were used in Gene Ontology analysis as described below (Panther).

Bioinformatic analysis

RNA-Seq data processing

Quality assessment and trimming of the de-multiplex reads was carried out using FastQC and cutadapt respectively (http://www.bioinformatics.babraham.ac.uk/projects/fastqc(1). The trimmed short reads were mapped to the human reference genome (hg19) using TopHat2 (version 2.0.12), a splice-aware mapper(2). Uniquely mapped reads were counted using HTSeq (version 0.6.0)(3) and the relative transcript abundance determined using DESeq2 (version 1.6.3)(4).

Other datasets

To compare the levels of transcripts in the yolk sac and placental villi with those in the liver, lung and kidney processed RNA-Seq data was down loaded from the EBI Expression Atlas (version: 0.1.4-SNAPSHOT, experiment E-MTAB-513 http://www.ebi.ac.uk/gxa/experiments/E-MTAB-513?ref=aebrowse).

To compare with 27 other tissues processed RNA-Seq data was downloaded from Expression Atlas - E-MTAB-1733 (http://www.ebi.ac.uk/arrayexpress/experiments/E-MTAB-1733)(5).

GO analysis

The most abundant 400 transcripts in the tissues of interest were identified (ranked by mean RPKM) and gene ontology analyzed by using Panther(6) (http://pantherdb.org release 20160715). The complete Panther data base was used to include the mitochondrially encoded transcripts. The Biological Process, Molecular Function and Cellular Component were all analyzed. The reference sets were all Homo sapiens or Mus musculus genes (known in Panther) and the Bonferroni correction for multiple testing was applied in all cases. Chord plots of enriched GO terms (Figure 1 for example) were generated in R using GOplot (v 1.0.2. available from CRAN) (7).

Venn diagrams

Venn diagrams were generated in R (vennCounts and VennDiagram, within the Bioconductor limma package) and the p-values for the observed overlaps calculated with SuperExactTest by Wang et al 2015 (v 0.99.2, available from CRAN)(8).

Transcription factor binding sites

Within the most abundant 400 transcripts in the human yolk sac, 19 genes are annotated as "regulation of transcription, DNA-templated" (GO:0006355) including several transcription factors (ATF4, FOS, FOSB, JUN, JUNB and JUND XBP-1 and BHLHE40). In the mouse 8 genes were similarly annotated (Table S3.)

The correlation (Spearman) between each of the transcripts encoding these transcription factors and all other yolk sac transcripts (rpkm >15) was calculated. The transcripts with rho >0.8 were identified. Over-represented transcription factor binding motifs in these genes were identified (JASPAR2016 and TFBSTools). Candidate motifs recognized by the transcription factors (or family for JUN and FOS) were enriched in the 1kb and 5kb upstream of the TSS of the highly correlated genes. Regions of interest were extracted using Biostrings and transcription factor binding motifs identified using JASPAR2016 and TFBSTools, all in R. P values were corrected for multiple testing(9) (Datasets S7 & 8).

Similar analysis was carried out for the mouse data in which 8 genes were annotated as above. Atf4 was the only transcription factor with a binding motif in the JASPAR2016 database. Candidate motifs recognized by Atf4 were enriched in the 1kb and 5kb upstream of the TSS of the highly correlated genes (Datasets S9 &10).

Other manipulations were carried out in R (BiomaRt, dplyr and made4) and Excel.

Results

Sample quality control

Hierarchical clustering of the yolk sac reads indicated that 1 sample did not cluster with any other sample and formed a separate branch. This sample had extremely high levels of villous specific transcripts (CGA, CGB, CGB5, CGB8, PSG1, PSG2, PSG3, PSG4, PSG9, LEP, KISS1 and CSH2; for example CGA \sim 73,000 vs a mean of \sim 320 for the other samples). It is therefore likely that the sample is contaminated with villous material. This sample was removed and the level of these transcripts examined in the remaining samples. Two additional samples had markedly different levels of these transcripts and so were also excluded from subsequent analysis.

RNASeq data from the first trimester villus samples were similarly processed and 1 sample clustered alone as a well separated branch. This sample had higher levels of decidual transcripts (such as IGF2, IGFPB4, IGFPB5, IGFPB7 and low levels of villous transcripts (CGA, CGB, CGB5, LEP and KISS1). This sample was therefore excluded from subsequent analysis.

A similar approach was adopted with the mouse yolk sac samples and 1 sample clustered as a well-separated branch. This sample had much lower levels of the fetal hemoglobin transcripts (Hbb-y, Hbb-bh1 and Hba-x) whereas these were the most abundant in all the other samples. This sample was therefore excluded from subsequent analysis.

Mitochondrial transcripts

Various enriched GO terms are associated with mitochondrial activity ("mitochondrial respiratory chain complex", and "mitochondrial ATP synthesis coupled electron transport" for example). This is consistent with the previously reported high density of mitochondria in the yolk sac(10). We calculated the proportion of transcripts present that were encoded by the mitochondrial genome (Figure S1). The fractions for the yolk sac and first trimester villi are slightly above the average and these data are consistent with that showing the kidney and heart had the highest mitochondrial fractions(11).

Figure S1. The fraction of mitochondrially encoded transcripts in each tissue type is the proportion mitochondrial genes over the sum over all RPKM values in that tissue

Mitochondrially encoded transcripts directly involved in the electron transfer chain and (or) energy generation were highly correlated with each other (Spearman's rho >0.71) and not with other transcripts in the most abundant 400 (Datasets S6). One other transcript was correlated with these (HNRNPA2B1). Other mitochondrially encoded transcripts (12S and 16S rRNA) were not similarly correlated.

Comparison with other tissues.

We used hierarchical clustering to examine the relationships among the expression patterns of the most variable transcripts detected at rpkm \geq 0.1 in the yolk sac, placental villi and a range of other tissues. As previously reported, clustering recapitulated the origin of the tissue(5). The first trimester placental villi and the term placental samples clustered together with the yolk sac cluster forming an adjacent branch (Figure S2). This indicates the first and third trimester villi have similar transcript profiles and the 3 extra-embryonic tissues are more similar to each other than to the other tissues examined. The yolk sac samples form 2 closely Figure S1. The fraction of mitochondrially encoded transcripts in each tissue type is the proportion mitochondrial geness over the sum over all RPKM values in that tissue type is the proportion mitochondrial geness over t

a marked difference in the level of transcripts encoding ALB, AFP, HBE, HBZ. It is known from previous studies that the size of the yolk sac declines from a maximum diameter of ~6mm at approximately 10 weeks gestation (12). The gestational age of the samples used in this study is only available for a subset of the samples. However, for those where this is known the samples obtained earlier in gestation (7+0, 8+2, 9+0 and 9+2) are on a separate branch to the known later sample (11+0). Male and female samples are distributed evenly across both branches.

Figure S2. Hierarchical clustering of multiple tissues based on the 500 transcripts with the

Figure S3 a & b.

Chord plots illustrating the proteins annotated with the GO molecular function and cellular

Summary of the number of RNASeq reads (single-end 50base) obtained for the human and mouse yolk sacs and the first trimester human placental villi.

Proteins in **bold** are also reported in human serum

Categorization of the proteins detected in the human coelomic fluid and comparison with serum proteins

Table S3

Human

Mouse

Table of the genes annotated with the GO term "regulation of transcription, DNA-templated" (GO:0006355) in the most abundant human and mouse yolk sac transcripts. GO evidence codes (go_linkage_type) are also provided (see http://geneontology.org/page/guide-goevidence-codes)

Table S4

Over represented GO terms shared among the 400 most abundant transcripts in the human, mouse and chicken yolk sacs. BP, Biological process; MF molecular function and CC, Cellular component

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