Two distinct waves of transcriptome and translatome remodelling drive germline stem cell differentiation

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Abstract

The process of stem cell differentiation is complex and yet reproducible, relying on the tight control of gene expression. However, a full understanding of the gene expression changes driving fate transitions has been hindered because stem cells are sparsely distributed in tissues. Here, we have overcome these technical limitations to perform RNA-seq and Ribo-seq at high temporal resolution during Drosophila female germline stem cell differentiation in vivo. While the data uncovers the extensive remodelling of both the transcriptome and translatome, we observed 3-fold more changes in translation compared to mRNA level. Furthermore, changes in mRNA level were frequently buffered by changing translation efficiency to stabilise the final rate of translation. Contrary to the expected cumulative changes, following gene expression throughout differentiation reveals two distinct waves of remodelling at both the transcriptome and translatome level. Altogether, our data provide a systematic and genome-wide roadmap of the changes in gene expression occurring during the differentiation process in vivo.
Introduction

Adult stem cells divide repeatedly, replenishing the stem cell population while producing daughter cells that undergo a change in fate and differentiate to specific cell types. The initial fate change is often driven by a master differentiation factor, which may be either asymmetrically segregated to one daughter cell, or upregulated in response to exclusion from a stem cell niche\(^1,2\). Many master differentiation factors have been identified by their mutant phenotypes: blocking normal differentiation. However, understanding the gene expression changes downstream of the master regulator is much more complex and experimentally challenging, especially in adult stem cells such as in the hematopoietic system, gut and skin\(^3\)–\(^6\). Adult stem cells are found in small numbers and are often quiescent for long periods of time, making it difficult to purify different stages of differentiation from living organisms, or to culture them in vitro.

Drosophila ovarian germline stem cells (GSCs) divide throughout adulthood to produce the oocytes, and are one of the best characterised in vivo stem cell systems\(^7\). Germline stem cells (GSCs) are located at the anterior of the ovary in a structure called the germarium, which houses the entire differentiation process\(^8\) (Figure 1A). GSCs are maintained in a stem cell niche, receiving Dpp/BMP signalling from the somatic cap cells. With each division, one cell is retained in the stem cell niche, while the other is normally excluded from the niche and initiates differentiation. The differentiating daughter cell undergoes four mitotic divisions, with incomplete cytokinesis, resulting in a 16 cell cyst (16cc) interconnected by a cytoplasmic bridge and a structure called the fusome. One of the 16 cells is selected to be the oocyte and meiosis is initiated, while the other 15 cells become nurse cells. Together with surrounding somatic epithelial cells, the 16cc will then form an egg chamber, beginning the process of vitellogenesis to produce the egg.

The master regulator of differentiation in GSCs is bag-of-marbles (bam)\(^9\). Dpp signalling from the somatic niche represses bam transcription in GSCs, so when a daughter cell is excluded from the niche, bam is upregulated and drives initiation of differentiation\(^10\). bam mutants cannot differentiate and so mutant ovaries accumulate GSC-like cells, the phenotype for which the gene is named. While the regulation of bam transcription has been extensively studied, its downstream effects are less clear. Most of the current understanding was brought about by the study of mutants affecting either stem cell self-renewal or differentiation. In the GSC, Pumilio (Pum) and Nanos (Nos) were shown to act together to repress translation of genes required for differentiation\(^11,12\). Acting with Benign gonial cell neoplasm (Bgcn), Bam is thought to target Pum and negatively regulate the Pum/Nos complex, allowing translation of the differentiation genes\(^13\). Downstream of that, and by means yet to be understood, the RNA binding protein (RBP) Rbfox1 is upregulated during the 4cc to 16cc stages. Rbfox1 was shown to inhibit Pum translation, consolidating the differentiation process\(^14\). Despite the identification of such players, many of which are RBPs, the cohorts of genes regulated by these and other factors during differentiation are unknown. Later during vitellogenesis, translation regulation plays another well-studied role in repressing premature production of protein from maternally deposited transcripts in the oocyte\(^15\).

While the understanding of the regulatory mechanisms driving GSC differentiation is limited, a myriad of cellular changes have been identified from GSC to 16cc. The most obvious cell fate change is a switch from complete cytokinesis in the GSC, which depends on ribosome biogenesis\(^16\), to incomplete cytokinesis in differentiating cysts via decay of the ESCRT-II machinery\(^17\)–\(^20\). Ribosome biogenesis is high in GSCs and is suppressed during differentiation\(^21\),\(^22\). At the same time, we have reported that global translation levels are upregulated in early differentiating cells compared to GSCs\(^16\) and this uncoupling is necessary for differentiation\(^23\). During differentiation, the microtubule network reorganises such that the minus ends orient to the fusome and then the future oocyte\(^24\), leading to the oocyte accumulating markers such as oskar (osk) mRNA\(^25\), the oo18 RNA-binding protein (Orb)\(^26\) and mitochondria\(^27\). The mitochondria also undergo morphological changes, increasing the number of cristae during differentiation through the dimerisation of ATP synthase\(^28\). Homologous chromosomes are unpaired in GSCs and undergo pairing from the 2cc to 8cc stages such that at the end of differentiation at the 16cc stage, chromosomes are paired in time for the initiation of meiosis\(^29\). These examples illustrate the extent and complexity of cellular changes during GSC differentiation but the regulatory processes directing them at a gene expression level are unknown.

Available methodology has limited the study of gene expression regulation during GSC differentiation.
Genome-wide screens have identified many genes with a role in GSC differentiation and oogenesis\textsuperscript{16,30}. Complementary low-throughput imaging-based methods can provide insight into some of the gene expression changes characterising GSC differentiation. Most recently, single cell RNA sequencing has been performed on whole ovaries by several groups\textsuperscript{31–34}. These studies successfully identify mRNA expression signatures between undifferentiated germline cells, immature and mature nurse cells and oocytes, as well as many different somatic cell types of the ovary. Pseudotime analysis has been used to subcluster germline cells but, due to the low depth of existing single cell technologies, present a superficial view of the transcriptome. Furthermore, it is not yet possible to assess translation regulation at a single cell level. Therefore, a genome-wide assessment of the transcriptome and translatome with high temporal resolution throughout differentiation of any in vivo stem cell system has not been achieved, hindering our understanding of the complex changes happening downstream of master regulators.

We have leveraged established genetic tools to synchronise GSC differentiation in vivo\textsuperscript{35–37}. In this system, we make use of a bam mutant to generate ovaries containing large pools of undifferentiated GSC-like cells. A heat shock promoter is then used to induce Bam expression and drive synchronous entry into differentiation. Ovary isolation at given time points after heat shock produces a sample enriched for GSCs or cysts at a given stage of differentiation. This protocol allows collection of sufficient material for genome-wide experiments including RNA-seq (measuring the transcriptome) and Ribo-seq (measuring the translatome) at high temporal resolution through GSC differentiation (6 time points from GSC to 16cc). We have extensively validated this method against wild type GSC differentiation using imaging-based methods. We find that the transcriptome undergoes a more significant transformation than expected, but nevertheless, the translatome undergoes 3-fold more changes than the transcriptome. Rather than a continuum of accumulating changes throughout differentiation, our data reveal two waves of global gene expression change at the level of both the transcriptome and translatome. The early wave upregulates cell cycle genes and downregulates translation-related genes, while the late wave upregulates genes involved in terminal differentiation (oogenesis and meiosis). In addition to uncovering the unexpected gene expression path driving germline stem cell differentiation, these datasets provide a framework for the remodelling of the transcriptome and translatome that happen during differentiation - a useful resource for developmental biologists.

Results

Establishing a robust protocol to synchronise GSC differentiation in vivo

Female bam mutant flies accumulate GSC-like cells that fill their ovaries, and these samples have been used previously to examine the transcriptome and translatome of the GSC\textsuperscript{38–40}. To measure gene expression changes with high time resolution during differentiation, we took inspiration from methods that induce germline differentiation by coupling a bam mutant with a transgene that can drive differentiation through brief restoration of Bam via a heat shock promoter (bam\textsuperscript{−/−},hs-bam)\textsuperscript{36,37,41,42}. Previous studies in the adult ovary have found that introduction of Bam via a heat shock promoter can drive differentiation in bam\textsuperscript{−/−} mutants such that ovaries contain well-formed egg chambers after 6-8 days\textsuperscript{35}. Therefore we asked whether shorter time points after Bam induction in a bam\textsuperscript{−/−},hs-bam ovary could be used to collect samples at different stages of differentiation (Figure 1A, Materials and Methods). We optimised the heat shock protocol to maximise the number of GSC-like cells entering differentiation, while minimising fly death due to heat stress.

Confocal microscopy was used to assign time points representing each stage from undifferentiated germ cells to the 16cc stage (terminal differentiation). GSC differentiation is characterised by four mitotic divisions with incomplete cytokinesis, leading to cyst cells joined by the fusome. We used the branching of the fusome (stained with alpha-spectrin)\textsuperscript{33,44} to observe the progression of differentiation at each time point after heat shock (AHS) (Figure 1B). Based on this branching, we assigned each time point to an approximate stage of differentiation: ‘no HS’ (bam mutant), 5 hr (cystoblast, CB), 9 hr (2cc), 18 hr (4cc), 28 hr (8cc) and 38 hr (16cc) AHS (Figure 1A). This time course provides a framework for studying GSC differentiation, but it is important to note that we observed significant heterogeneity in cyst stages, which increased at later time points.
To visualise the kinetics of Bam expression induced by our heat shock protocol, we performed Western Blot analyses to measure Bam protein level in whole ovaries throughout the time course. The results revealed that Bam protein was strongly induced at 0.5 hr AHS and then declined to levels similar to those observed in the ‘no HS’ samples by 18 hr AHS (Figure 1C). As a control, we measured Vasa (a germline marker) levels across the time course and observed no changes.

To monitor the variation in Bam induction between cells that is not captured by Western blot analyses, we used immunofluorescence (IF). In wild type ovaries, Bam protein is not expressed in the GSC, but is upregulated during early differentiation (Figure 1D). In our synchronised GSC system, Bam is not detected before the heat shock, but is rapidly upregulated in most cells at 0.5 hr AHS. Bam levels decrease dramatically by 18 hr AHS, and as observed by Western Blot, Bam is not detectable by 28 hr AHS using IF. The imaging
analysis revealed that the loss of Bam protein is variable: at 9 hr AHS, we observed a wider range of Bam intensity between differentiating cysts in comparison with 0.5 hr or 5 hr AHS. While the basis of such variability has not been determined, it is possible that this explains the increasing heterogeneity of cyst development we observed by 38 hr AHS (Figure 1B).

To investigate whether normal differentiation proceeded as expected, we examined a marker defining the final stages of differentiation and early stages of egg chamber development. During normal differentiation, Orb is concentrated into the future oocyte at the 16 cell cyst stage. IF revealed that Orb does not concentrate into a single cell of the 16cc by 38 hr AHS, but most egg chambers contain a single Orb positive cell at 5 days AHS (Figure 1E). Furthermore, and as previously shown, females treated with our heat-shock protocol laid eggs that hatched and produced adult offspring, confirming that normal germline development could be fully recapitulated.

Transcriptome measurements in synchronised GSCs reflect wild type differentiation

To examine mRNA expression during differentiation in synchronised GSCs, we performed RNA-seq on RNA-depleted RNA extracted from bam−/−,hs-bam ovaries collected at specified time points AHS (Figure 1A). We generated RNA-seq libraries from ovaries with ‘no HS’ (bam mutant) and 5 hr (CB), 9 hr (2cc), 18 hr (4cc), 28 hr (8cc) and 38 hr (16cc) AHS. Comparison between biological replicates showed that the libraries are highly reproducible (R²=0.97). Using a low threshold of 2 FPKM (Fragments Per Kilobase per Million mapped reads), we detected a total of 8190 genes expressed in at least one time point.

To validate that the transcriptome measurements in the bam−/−,hs-bam time course reflect mRNA accumulation during wild type differentiation, we used single molecule fluorescence in situ hybridisation (smFISH) to visualise mRNA transcripts in wild type germaria (Figure 2A, S1A). Marking the cell boundaries (phalloidin, F-actin) and the spectrosome and fusome (alpha-spectrin) enabled us to identify all the stages of differentiation in wild type ovaries. vasa mRNA is detected at ~ 170 FPKM throughout the RNA-seq time course, while aubergine (aub) mRNA is detected at a lower, but constant, level (~ 80 FPKM). Accordingly, using smFISH, we observed higher levels of vasa mRNA transcripts in the cytoplasm of germ cells at all differentiation stages (Figure 2Ai), compared to aub mRNA (Figure 2Aii). matrimony (mtrm) mRNA presented levels below the 10 FPKM threshold until the 28 hr AHS time point, with a further increase at 38 hr AHS. This pattern of upregulation is also observed in wild type differentiation, with mtrm mRNA only observed in 8cc and 16cc (Figure 2Aiii). bgcn displays the opposite mRNA expression pattern, with declining mRNA levels in both the synchronised differentiating GSCs and wild type differentiation (Figure 2Aiv).

Notably the synchronised differentiating GSC dataset recapitulates recently published mRNA expression patterns for genes such as CG32814, which is reported to be downregulated after the GSC and blanks, which is downregulated at the 8cc stage (Figure S1A). Collectively, validations by smFISH indicate that the RNA-seq data in synchronised differentiating GSCs mirrors what is observed in wild type ovaries both in terms of mRNA levels and changes during differentiation. The smFISH experiments suggest that an expression level in the RNA-seq below 10 FPKM corresponds to a negligible number of mRNA transcripts per cell, so we set this as the threshold for further analysis. In our RNA-seq data, 6532 genes were expressed at > 10 FPKM in at least one time point.

The protocol used to drive Bam expression is also expected to induce acute changes in the expression of the heat shock responsive genes. Transcript level changes of the heat shock responsive genes are characterised by their transient nature, quickly dampening over short periods after the stimulus is removed. To identify genes with acute changes in mRNA level upon heat shock treatment, we generated RNA-seq libraries from ovaries at 0.5 hr AHS. 68 genes underwent a significant > 3-fold change in transcript level in the 0.5 hr AHS time point compared to ‘no HS’ (with > 10 FPKM expression in at least one of these time points). Most of these genes recovered towards the ‘no HS’ level during the time course, but at differing rates (Figure S1B). Gene ontology analysis showed that this cohort of genes were highly enriched for heat response and protein folding terms (Figure S1C). Therefore, these genes, along with an additional 20 contaminant genes (encoding proteins of the chorion and vitelline membrane, which are produced by somatic cells), were excluded from downstream analysis (Supplementary Table 1). This filtering resulted in a dataset of 6444 genes expressed during the time course (Figure 2B, Supplementary Table 2).
Figure 2 – RNA-seq in synchronised differentiating GSCs recapitulates changes in gene expression during normal differentiation

(A) single molecule fluorescent in situ hybridisation (smFISH) was used in wild type germaria (right) to validate the RNA expression measurements in the synchronised differentiating GSC RNA-seq dataset (left). Wild type germaria are stained for DNA (Hoechst, blue), actin (phalloidin, gray), fusome (α-spectrin, yellow), and mRNA transcript of interest (smFISH, red and grayscale): (i) vas, (ii) aub, (iii) mtrm and (iv) bgcn. Scale bars are 15 µm. (B) RNA-seq in the differentiating GSCs detected 8177 genes expressed at least 2 FPKM at one time point (excluding the 0.5 hr heat shock sample), and 6535 of these were expressed to at least 10 FPKM. 90 genes were excluded due to a strong heat shock effect or chorion contamination. (C) Heat map showing fold change in gene expression (RNA-seq), compared to ‘no HS’. Each row represents one of the 548 genes that exhibit a significant 1.6-fold change in expression between two time points. Genes were grouped by fold change in mRNA expression, starting with the final time point. gold = two-fold increase compared to no HS, black = no change, cyan = two-fold decrease. (D) Number of genes with a significant 1.6-fold change at each time point compared to ‘no HS’. Gold = upregulation, cyan = downregulation. (E) Illustrating the overlap between genes with significant up or down regulation at 9 hr AHS and 38 hr AHS. Orange and blue represent genes with changing expression at 9 hr AHS, and these are followed to 38 hr AHS. Additional genes with changing expression at 38 hr AHS are coloured pink and green.
GSC differentiation involves dynamic changes in the transcriptome

Having established that the synchronised GSC RNA-seq dataset parallels what is observed during wild type differentiation, we aimed to identify the set of genes for which mRNA levels change during GSC differentiation. To do so, we performed differential expression analysis and found 548 genes with a significant (p < 0.05) and greater than 1.6-fold difference in mRNA level between any two time points (in which the expression was > 10 FPKM in one of the compared time points; and excluding the 0.5 hr AHS sample) (Figure 2C). We counted the number of genes with a significant >1.6 fold change in RNA level at each time point relative to ‘no HS’. At any given time point there are a maximum of 128 changed genes compared to ‘no HS’ and in total there are roughly an equal number of upregulated and downregulated genes. However, we found that most mRNA downregulation (65%) is observed at the 5 hr and 9 hr AHS timepoints (CB and 2cc), while most upregulation (64%) is observed at the 28 hr and 38 hr AHS time points (8cc and 16cc) (Figure 2D). The 113 genes changing at 38 hr AHS represent only 21% of the genes with a significant fold change between any two time points (represented in Figure 2C), which is inconsistent with a model of gradual accumulating changes from GSC to 16cc. Indeed, we found that most genes (84%) that are up or downregulated at 9 hr AHS (2cc) returned to ‘no HS’ levels by 38 hr AHS (16cc) (Figure 2E). In contrast, the majority (82%) of mRNA expression changes at 38 hr AHS are contributed by genes which exhibited no change at 9 hr AHS. We performed Principal Component Analysis (PCA) to understand the variance over the time course (Figure S2A). PC1 explains ~37% of the variance and forms three clusters: 5 hr AHS and 9 hr AHS, no HS and 18 hr AHS, 38 hr AHS and 28 hr AHS. The 9 hr AHS and 38 hr AHS samples are the most separated, illustrating an inflection point around 18 hr AHS, which is most similar to no HS in PC1. PC2 explains ~28% of the variance and groups the mid samples (9 hr AHS, 18 hr AHS and 28 hr AHS) closer together. This data reveals that there are dynamic changes in the transcriptome throughout differentiation, potentially driving different cellular changes at different stages of differentiation.

To further characterise the observed changes, we asked which genes contribute to each wave of mRNA expression change, and performed gene ontology analysis. We considered genes with mRNA level changes at 9 hr AHS (2cc) and at 38 hr AHS (16cc) in comparison to ‘no HS’ (Figure S2B). We found that the group of genes upregulated at 9 hr AHS was enriched for genes involved in DNA replication and the cell cycle (Figure S2C), while downregulated genes were enriched for genes involved in regulation of lipid storage. At 38 hr AHS the upregulated genes were enriched for annotations of the polar granule (GOterm: P granule) (Figure S2C), which are assembled during oogenesis47, with no significant enrichment for downregulated genes. Indeed, the direct comparison of the final time point, 38 hr AHS, to ‘no HS’, revealed that the upregulated group includes genes known to play a role in meiosis (mtrm, c(3)g, orb, corolla) and oocyte development or early embryogenesis (osk, brul, dhd, png, alphaTub67C), while the downregulated group of genes includes bgn, blanks, Rpl22-like and eIF4E3.

With the advent of single cell sequencing technologies, other studies have used single cell (sc)RNA-seq analysis to study adult Drosophila ovaries in recent years31–33. To determine similarities between the datasets, as well as to assess the depth and resolution provided by different approaches, we compared the data generated on bulk RNA-seq from the synchronised GSCs to published single cell data. In each of the scRNA-seq studies, germ cells formed two clusters: one expressing early differentiation genes such as bam and corolla (“before 16cc”), and one expressing late differentiation genes such as osk, orb and dhd (“after 16cc”). The six time points in our synchronised differentiating GSC dataset are contained within the ‘before 16cc’ cluster, providing increased temporal resolution of gene expression during differentiation. Rust et al. and Slaidina et al. used pseudotime analysis to reveal more detailed dynamics of gene expression change during differentiation, identifying transcriptional signatures for different stages of germline development. We examined the expression of different groups of genes identified by Rust et al., in our synchronised differentiating GSC dataset: marker genes for different time points (Figure S3A), spindle genes expressed during the mitotic cyst division stage (Figure S3B) and synaptonemal complex genes and additional candidates expressed at the 8cc and 16cc stages (Figure S3C). We found that the expression patterns of those genes in our time course were highly consistent with the scRNA-seq. Using pseudotime analysis from GSC to early egg chambers the Rust et al. dataset identified 1930 candidate genes with changing expression, while 813 genes were identified in the Slaidina et al dataset. For comparison, without thresholding for expression...
Figure 3 – Ribo-seq in synchronised differentiating GSCs recapitulates changes in translation during normal differentiation

(A) After filtering, 6444 genes are expressed at >10 FPKM in at least one time point by RNA-seq. 548 of these genes show a >1.6-fold change between any two samples. Of these 548 genes, 353 also show a 1.6 fold change in Ribo-seq reads between two samples. 1360 genes show a >1.6-fold change in the Ribo-seq with no significant change in the RNA-seq. (B) bam transcript level and translation during the time course from RNA-seq and Ribo-seq data (FPKM). solid line = RNA-seq, dashed line = Ribo-seq. (C) RNA-seq and Ribo-seq in the synchronised GSC system recapitulates previously characterised regulation of osk and nanos. solid line = RNA-seq, dashed line = Ribo-seq. IGV tracks for each time point for RNA-seq (red) and Ribo-seq (blue). (D) GFP-tagged FlyFOS constructs (modulo, dodeca satellite-binding protein 1, calcium-independent receptor for a-latrotoxin) stained for gfp smFISH (magenta and individual grey scale) and GFP protein (green, and individual grey scale), f-actin (phalloidin, grey scale) and fusome (α-spectrin, yellow). Lines were selected because they have a similar level of mRNA expression but differing levels of translation (graphs, left, solid line = RNA-seq, dashed line = Ribo-seq). Scale bars are 15 µm.
level or fold change, we find 3457 genes with a significant (p < 0.05) change in expression during our time course of just the differentiating stages of the germ cells. In conclusion, our dataset is consistent with published scRNA-seq, while providing substantial increased detection depth.

Riboseq captures translation dynamics during GSC differentiation

While RNA-seq revealed the mRNA transcripts available at each stage of differentiation, mounting evidence indicates that transcripts for different genes are not equally translated48. Translation regulation plays a significant role during GSC differentiation - indeed the master differentiation factor Bam is itself a regulator of translation13. Despite this, the dynamics of translation changes during stem cell differentiation are unknown, largely due to the large amount of material required to analyse translation genome-wide. Importantly, translation cannot currently be studied genome-wide in single cells. To circumvent these difficulties, we applied the bam-3,hs-bam system to obtain enough starting material to perform Riboseq at the same time intervals as the RNA-seq dataset. Even using the synchronisation approach, the amount of tissue was limited compared to previous Riboseq experiments in Drosophila S2 cells49,50, embryos51, oocytes52 or ovaries53, so we first performed quality control analysis on the Riboseq libraries (Figure S4). Comparisons between biological replicates revealed that the data is highly reproducible (R2 > 0.97). As expected from high quality Riboseq data, we observed that the majority of the reads (58-70%) showed the expected ribosome footprint (28-32nt long), overwhelmingly mapped to the sense direction of the CDS (>90%) of 5’ UTR (~8.9%) and showed a strong 3-nucleotide P-site periodicity. This benchmarking confirms that our Riboseq is of high quality despite the limitations in input material.

We assessed the translation of the 6444 genes that we had determined to be expressed by RNA-seq (Figure 3A, Supplementary Table 3). Of these, 43 genes showed heat shock effects in the Riboseq (> 3-fold change in the 0.5 hr AHS sample compared to ‘no HS’). Using a threshold of 10 FPKM in the Riboseq, we found that 5922 of the expressed genes were translated at some point during our time course, while 479 genes were not translated at any time point (< 10 FPKM).

Just as we validated the RNA-seq datasets against wild type differentiation using smFISH, we aimed to validate the Riboseq through independent approaches. We first compared the Riboseq reads for bam to the results we obtained by Western blot (Figure 1B). From the Riboseq, we found that bam translation peaks at 0.5 hr AHS, but declines gradually (Figure 3B), which matches the Western blot. We then focused on hallmark genes that have been previously characterised as being translationally regulated during GSC differentiation. For instance, osk mRNA is not expressed until the late stages of GSC differentiation, but remains translationally repressed in differentiated cells until much later in oogenesis, becoming translated during egg chamber development25,54.55. In agreement, the analysis of the RNA-seq data revealed a progressive increase in the level of osk mRNA from 18-38 hr AHS (4-16cc), while the Riboseq showed no osk mRNA was engaged with ribosomes throughout the time course (Figure 3Ci). On the other hand, transcriptional reporters suggested that the nos gene is transcribed throughout GSC differentiation, while the protein level decreases during the CB to the 4cc stage13. The Riboseq data revealed a sharp decrease in nos translation at the 5 hr and 9 hr AHS timepoints (CB-2cc), but surprisingly, the mRNA level mirrored this change, which together with constant transcription implies decreased mRNA stability (Figure 3Cii). Both the RNA-seq and Riboseq showed full recovery of nos by 28 hr AHS (8cc).

While our datasets recapitulate examples of translation regulation already shown in the literature, we wanted to validate the Riboseq in an unbiased manner. To do so, we selected genes with similar RNA expression levels but different Riboseq read levels and procured FlyFOS GFP-tagged transgenic lines. The FlyFOS transgenes are expected to recapitulate the expression of the endogenous gene, as they were designed with the goal to include all regulatory sequences existing on endogenous genomic loci56. We stained germaria with gfp smFISH to measure RNA level, and GFP protein to measure the level of each fusion protein (Figure 3D), using the same gfp smFISH probes and anti-GFP antibody to minimise confounding detection effects. In agreement with the RNA-seq data, the smFISH showed similar levels of RNA for each gene. Moreover, mirroring the results from the Riboseq analysis, the GFP protein levels were very different for each gene (highest = mod, medium = Dpl, lowest = Cir). Therefore, we conclude that the Riboseq in synchronised differentiating GSCs reflects changes in translation during wild type differentiation. Riboseq data reports the rate of protein production during differentiation, providing more direct insight into the
changes in protein complement during differentiation compared with RNA-seq.

**Ribo-seq identifies global remodelling of translation during GSC differentiation**

We asked if changes in mRNA expression corresponded to changes in the Ribo-seq. 548 genes showed a >1.6-fold change in RNA level between any two time points, and 353 of these (64%) also showed a significant 1.6-fold change between two samples in the Ribo-seq dataset (Figure 3A). This implies that at least 36% of the transcriptomic changes were not reflected in the translatome within our time course. An additional 1360 genes showed a change in the Ribo-seq but not the RNA-seq data, revealing that these genes are regulated primarily at the level of translation. In total, 9% of
transcribed genes show a >1.6-fold change in RNA expression at some point over the time course, while 29% of translated genes show a >1.6-fold change in the Ribo-seq experiment. This result illustrates that a more extensive remodelling of the translatome than the transcriptome is at play during GSC differentiation.

1713 genes showed a >1.6-fold difference in translation level between any two time points (excluding the 0.5 hr AHS sample) (Figure 4A). Comparing the 38 hr AHS time point to ‘no HS’, 99 genes were upregulated and 165 were downregulated at the level of translation. Genes with increasing translation include orb, grk, moon, png, c(2)M, bru1, alphaTub67C, Rbp9. Genes with decreasing translation include bgcn, stau, Pxt. For some of these genes, the changes in translation reflect similar changes in mRNA level, e.g. orb and bgcn, but for others the Ribo-seq is not paralleled by the RNA-seq, implying that the latter group is regulated at the level of translation efficiency (Figure 5A). Similarly to osk, dhd is translationally repressed. On the other hand, wit is translationally upregulated such that Ribo-seq reads go up, while RNA-seq reads are stable during differentiation.

We determined the number of genes with >1.6 fold changes in the Ribo-seq at each time point relative to ‘no HS’ (Figure 4B). There were roughly twice as many downregulated genes as upregulated genes at each time point, with two waves of changes around an inflection point at 18 hr AHS. Similarly to the RNA-seq analysis, we found that the majority (88%) of genes that increase or decrease in translation in the early wave (9 hr AHS, 2cc), returned to ‘no HS’ levels by 38 hr AHS (16cc) (Figure 4C). The majority (77%) of translation changes in the late wave are contributed by genes that exhibited no change in the early wave. This pattern is reflected in PCA analysis of the Ribo-seq samples (Figure S5B). PCA1 explains ~34% of variance and shows the biggest difference between 9 hr AHS and 38 hr AHS, with ‘no HS’ and 18 hr AHS between the two. Indeed, the ‘no HS’ and 18 hr AHS samples are found close together in the PCA, illustrating that, similar to the RNA-seq, there is an inflection point of translation changes at 18 hr AHS.

We performed gene ontology analysis and found that, as in the RNA-seq, the first wave of upregulation was enriched for genes involved in cell cycle and DNA replication, as well as cytoskeleton organisation (Figure S5C, S5D). The early wave of downregulation in the Ribo-seq was enriched for genes involved in translation, primarily the ribosomal proteins. There was no significant enrichment in the second wave at the level of translation, implying that, as for osk, some of the upregulated transcripts that are enriched in the polar granules are translationally repressed at this stage.

**Translation efficiency is dynamically regulated during GSC differentiation**

RNA-seq measures gene expression changes at the mRNA level, which are often thought of as a proxy for consequential changes at the protein and cell fate level. Indeed, scRNA-seq has been broadly used as a readout for fate changes. Yet, this premise assumes no change in the translation efficiency (TE) of the transcripts. However, accumulating evidence indicates that changes in the translatome are a composite of differences in mRNA level and differences in TE of each gene. TE is a measure that is proportional to ribosomes per transcript, and can be calculated for each gene as Ribo-seq reads divided by RNA-seq reads. Looking at comparisons between ‘no HS’ and each subsequent time point, we observed 485 events with a >1.6-fold change in mRNA level, but only a third of these resulted in an equivalent >1.6-fold change in translation in the same direction at the same time point in the Ribo-seq. This reveals that the majority of changes in mRNA level occurring during GSC differentiation do not determine changes in protein production in those cells. On the other hand, there are an additional 1179 events of >1.6-fold change in translation between any time point and ‘no HS’ which do not show an equivalent change in mRNA level. Altogether, these findings suggest that regulation at the level of TE plays a prominent role during GSC differentiation.

To determine the regulation of TE, we used deltaTE to detect changes in TE throughout our time course compared to ‘no HS’. With a threshold of >1.6 fold change in TE (with p value <0.05 and FPKM of at least 10 in one of the compared RNA-seq samples), we identified 874 genes with changing TE during the time course (Figure 5A). While changes in the Ribo-seq are a good proxy of the outcome of protein production, studying the patterns of changes in TE could provide insight into regulation, unveiling patterns of coregulated genes. In this context, we asked if different sets of genes are regulated by mRNA level or translation efficiency. We looked at genes with a >1.6-fold change in translation (Ribo-seq) as well as a >1.6-fold change in either RNA-seq (mRNA level regulation) or TE (regulation at the level of TE) (Figure S6). We found
that cell cycle and DNA replication genes are enriched in the mRNA level upregulation group but not the TE upregulation group at 9 hr AHS, while translation genes are enriched in the TE downregulation group, but not the mRNA level downregulation group at 9 hr AHS. This suggests that the previously observed downregulation of ribosome biogenesis during GSC differentiation is primarily regulated at the level of translation efficiency. At 38 hr AHS the polar granule genes are enriched in the mRNA level group but not the TE group.

To understand the different types of gene regulation in more detail, we grouped genes into four classes: regulation only at the mRNA level (forwarded), only at the level of translation efficiency (exclusive), or by mRNA and translation efficiency changing in the same direction (intensified) or opposite directions (buffered).
We classified gene regulation into these four categories at each time point relative to the 'no HS'. Overall, we found that translation regulation contributed to the regulation of more genes than mRNA level, with the exception of the 18 hr AHS time point where 96% of regulatory events involved changing mRNA level. Intuitively, we had expected 'intensified' to be a large category, as it represents the most effective way to change protein expression. However, at each time point, only 4-10% of regulatory events are categorised as intensified. On the other hand, the 'buffered' category was larger, at each time point making up 14-56% of regulatory events. Together, these results reveal that changes in mRNA accumulation are a less consequential output than measurements obtained by Ribo-seq, and that translation regulation is the main step of gene regulation during germline stem cell fate transition.

**Discussion**

During differentiation, stem cell progeny undergo a fate change that can transform cellular morphology, metabolism and function. The reproducibility of development and homeostasis in general, and of germline stem cell differentiation in our particular case, suggests that the orchestration of changes in gene expression during fate changes is tightly regulated. Many key gene expression regulators have been identified in GSC differentiation, but their downstream effects on expression are poorly understood due to technical limitations of studying a small number of cells at specific stages of differentiation in a complete adult tissue. We have developed a protocol to synchronise GSC differentiation in vivo, allowing us to collect tissue samples representing the progression through differentiation that are amenable to in-depth, robust, genome-wide analysis. Indeed, this method generates sufficient material to perform RNA-seq and Ribo-seq at high resolution throughout GSC differentiation, providing a rich insight into the remodelling of the transcriptome and translatome during this cell fate change.

We performed benchmarking analysis to confirm the quality and reproducibility of our sequencing datasets, and thoroughly validated them against normal wild type differentiation in comparison to the published literature and using orthogonal imaging-based methods. The RNA-seq data of synchronised differentiating GSCs reproduces the gene expression patterns identified through scRNA-seq, and the increased depth of the data provided by RNA-seq allowed us to identify more differentially expressed genes in our dataset. More importantly, in our Ribo-seq experiment, we found that the translatome undergoes a more substantial transformation than the transcriptome during GSC differentiation, which could explain why it has been difficult to specifically distinguish signatures for each stage of differentiation from scRNA-seq data. Indeed, we found that during GSC differentiation there were 3-fold more genes undergoing changes in the translatome than in the transcriptome. While the role for translation control is not surprising, given that translation regulators have been shown to be drivers of GSC differentiation, the magnitude of the effect was not anticipated. Importantly, we found that changes in the RNA-seq were not predictive of changes in the Ribo-seq: just a third of the observed changes in the RNA-seq also show a concomitant change in the Ribo-seq. This finding is in line with previous reports that transcript level (as measured by RNA-seq) is not a good proxy for protein level. RNA-seq enables some insight into the regulatory processes upstream of protein production, but on its own provides limited explanation of the cell biological changes happening during cell fate transitions.

While we found hundreds of changes in gene expression at the level of the transcriptome and translatome, gene ontology enrichment analysis uncovered fewer enriched terms than we expected. This is likely because the activity of any particular pathway can be regulated by the level of one limiting component. However, we did identify that early in differentiation there was an enrichment for genes involved in the cell cycle and DNA replication. This upregulation may be important for driving the four mitotic divisions that occur during differentiation. Indeed, later in differentiation, this group of genes are no longer upregulated. It has been previously proposed that when the RNA-binding protein Bruno is expressed late in differentiation, it translationally represses mitotic factors leading to exit from the cell cycle. In contrast, our findings suggest that the cell cycle factors are regulated primarily at the mRNA level. During early differentiation, the downregulated genes in the Ribo-seq were enriched for terms involving the ribosome and translatome, and this was regulated by changing TE. This finding is consistent with previous literature that ribosome biogenesis is downregulated during differentiation and suggests that ribosome biogenesis is regulated at the level of translation. At late stages of differentiation, the transcriptome showed an upregulation of genes enriched in polar granules. This was not reflected in the
Ribo-seq, likely because many of these genes were shown to be translationally repressed at the early stages of oogenesis. At both the level of transcription and translation, we observed two distinct waves of gene expression changes: at both 5-9 hr and 28-38 hr AHS there were many changes relative to ‘no HS’, while at 18 hr AHS there is an inflection point which is most similar to the ‘no HS’ undifferentiated sample. Interestingly, the genes with changing expression in each wave had minimal overlap, suggesting that the differentiation to a new fate is characterised by two separate phases rather than by the gradual accumulation of changes. The inflection point at 18 hr AHS shows surprisingly few changes from the GSC-like ‘no HS’ sample, but has a very different forward trajectory of gene expression changes. The first wave of gene expression changes are driven by the introduction of Bam protein via the heat shock, and Bam levels have resolved to background levels by 18 hr AHS. However the cells at 18 hr AHS do not return to a stable GSC-like fate, and instead embark on the second phase of differentiation without further extrinsic stimulus. We hypothesise that changes inherited through the first phase, and the differences between the 18 hr AHS and ‘no HS’ sample are responsible for directing this intermediate stage into the second phase of differentiation. Regarding the direct comparison of expression between 18 hr AHS and the ‘no HS’ sample, two changes stand out: the loss of Bgcn and the upregulation of Rbfox1. Bgcn acts with Bam to antagonise the action of Pum, a translational repressor which blocks differentiation of the GSC. At 18 hr AHS neither Bam nor Bgcn is present, but Rbfox1 upregulation has been shown to prevent the reversion into a stem cell fate by repressing translation of pum. Rbfox1 expression is dependent on Bam and is thought to control the transition from the mitotic to post-mitotic stages of differentiation. We speculate that a second pulse of Bam at 18 hr AHS would have very different regulatory effects, due to the loss of Bgcn and presence of Rbfox1.

We analysed how mRNA level and translation efficiency each contribute to changes in translation during differentiation. Changing both the mRNA level and translation efficiency in the same direction has an amplifying effect, but unexpectedly, this ‘intensified’ category made up a minority of regulatory events. Indeed, when both mRNA level and translation efficiency acted on the same gene, they most frequently acted in the opposite direction (buffering the level of translation). The buffered category includes genes which are upregulated at the mRNA level, but remain translationally repressed, with two extreme cases being osk and dhd. In this way, buffering enables a preparatory transformation of the transcriptome, without immediately affecting the protein complement of the cell. Additionally, buffering could be a mechanism of modulating the translation efficiency that leads to a more precise regulation than mRNA level alone.

Altogether, we have uncovered two phases of gene expression change, which may first drive mitosis and the removal of the self-renewal fate, and subsequently promote terminal differentiation. Many adult stem cells undergo a transit amplifying stage to increase cell number before terminal differentiation. Therefore, the gene regulatory mechanisms uncovered here may be broadly applicable to these stem cell systems, including in the larval brain as well as intestine, skin and hematopoietic systems. Furthermore, the datasets presented here can also be interrogated at a single pathway or gene level to provide insight into a large diversity of biological questions during differentiation. In this way, it will be a valuable resource for stem cell and developmental biologists.

Acknowledgments
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Author contributions
TJS, JG and FKT conceived the idea and designed the experiments. TJS and JG performed the experiments. TJS, DG and FKT analysed the sequencing data. TJS and FKT wrote the manuscript.
Declaration of Interests
Authors declare no competing interests.

References


10.1126/science.abg2653.


Single-cell reconstruction of developmental trajectories during zebrafish embryogenesis. Science 360, eaar3131. 10.1126/science.aar3131.


**Materials and Methods**

**Resource availability**

**Lead Contact**

Further information and requests for resources or reagents should be directed to the lead contact Felipe Karam Teixeira (fk319@cam.ac.uk).

**Data and code availability**

RNA-seq data generated during this current study will be deposited at GEO and will be publicly available as of the date of publication. This paper does not report original code. Any additional information required to reanalyse the data reported in this paper is available from the lead contact upon request.

**Drosophila husbandry and genetics**

Unless otherwise stated, stocks and crosses were maintained on standard propionic food at 25°C for experiments. The *Drosophila melanogaster* stocks used were:

<table>
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<tr>
<th><em>Drosophila</em> stock</th>
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<th>Origin</th>
</tr>
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<tbody>
<tr>
<td>wild type</td>
<td>w[1118]</td>
<td>R. Lehmann lab</td>
</tr>
<tr>
<td>bam&lt;sup&gt;Δ&lt;/sup&gt;,hs-bam</td>
<td>w[1118]: nosP-GAL4-NGT40/CyO; bamΔ86, P[w+[+mC]=hs-bam.O]11d/TM6B</td>
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</table>

**Synchronising GSC differentiation**

Virgin females (genotype - w[1118]; nosP-GAL4-NGT40/CyO; bamΔ86, P[w+[+mC]=hs-bam.O]11d/TM6B) were crossed to male (genotype - w;; bamΔ86/Tm6C). Female F1s (genotype - w;nosP-GAL4-NGT40/++;bamΔ86, P[w+[+mC]=hs-bam.O]11d/bamΔ86) were collected overnight and fattened on yeast for three days. Flies were heatshocked in pre-warmed vials without yeast for 1 hour at 37°C, followed by 2 hours at 34°C. Flies were flipped to fresh vials with yeast, returned to 25°C and time after heat shock (AHS) was measured from this point.

**smFISH probe labelling**

smFISH probes (32 probes per gene) were designed using the Stellaris Probe Designer (Biosearch Technologies) and ordered as unlabelled DNA oligos. Labelling was done according to Gaspar et al., 2017<sup>69</sup>. Briefly, unlabelled ddUTP was conjugated to an ATTO dye NHS ester (ATTO565 or ATTO633, Atto-tec), then the labelled ddUTP was added to the 3’ end of each probe with terminal deoxynucleotidyl transferase. Probes were purified by ethanol precipitation.

**Antibody staining**

Ovaries were dissected in cold PBS and fixed for 25 minutes at room temperature in 4% formaldehyde in 0.3% PBSTX (0.3% Triton-X). Ovaries were washed with 3x 15 mins in 0.3% PBSTX ovaries were incubated in Block (0.2 μg/ul BSA in 0.3% PBSTX) for 1 hr at room temperature. Primary antibodies were added to Block at the appropriate concentration for incubation overnight at 4°C. The following day, washes and secondary antibody incubation were performed in Block, with the addition of Hoechst 33342 DNA stain (ThermoFisher Scientific) in one wash step.
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<tr>
<th>Antibody</th>
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</tr>
</thead>
<tbody>
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<td>Alpha-spectrin (mouse)</td>
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<td>DSHB Cat#3A9 (323 or M10-2)</td>
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<td>Vasa (rabbit)</td>
<td>1:1000</td>
<td>Ruth Lehmann</td>
</tr>
<tr>
<td>Bam (mouse)</td>
<td>1:20</td>
<td>DSHB bam</td>
</tr>
<tr>
<td>Orb (mouse)</td>
<td>1:200</td>
<td>DSHB 4H8</td>
</tr>
<tr>
<td>GFP Booster ATTO 488</td>
<td>1:500</td>
<td>Chromotek (gba488-100)</td>
</tr>
</tbody>
</table>

**smFISH**

Ovaries were dissected and fixed as above for IF. After 3x washes in 0.3% PBSTX, samples were transferred to Wash buffer (2X saline sodium citrate (SSC), 10% deionised formamide in nuclease-free water) for 10 mins at room temperature. smFISH probes, primary antibodies and phalloidin (Alexa Fluor 405 or 488 Phalloidin, ThermoFisher Scientific) were diluted in Hybridisation buffer (2X SSC, 10% deionised formamide, 20mM vanadyl ribonucleoside complex, 0.1 mg/ml BSA, competitor (1:50 dilution of 5 mg/ml E.coli tRNA and 5 mg/ml salmon sperm ssDNA) in nuclease-free water). Ovaries were incubated in Hybridisation buffer at 37°C overnight (less than 16 hours). Ovaries were washed 3x 15 mins in Wash buffer, then incubated with secondary antibodies in Wash buffer for 2 hours at room temperature. Ovaries were finally washed in Wash buffer, with the addition of Hoechst in one wash step.

**Imaging**

Samples were mounted in VectaShield mounting media (Vector Laboratories). Images were acquired on a Leica SP8 confocal microscope with a 20x dry objective or 40x oil objective. Image processing was using Fiji.

**Western blot**

30 pairs of ovaries were collected for each sample and lysed in Laemmli buffer with B-mercaptoethanol. Ovaries were homogenised with an electric pestle then incubated at 95°C for 5 min. Samples were run on a Novex Value 4-12% Tris-Glycine Mini Protein Gel. Proteins were transferred to a PVDF membrane by wet transfer. Blots were blocked with 5% skimmed milk in TBST (TBS with 1% Tween) for 1 hr at room temperature. Primary antibody incubation was at 4°C overnight, then blots were washed and incubated with secondary antibodies in TBSTS (TBST with 0.01% SDS) for 30 mins at room temperature. Blots were imaged with a Licor Odyssey imager.

<table>
<thead>
<tr>
<th>Antibody</th>
<th>Concentration</th>
<th>Origin</th>
</tr>
</thead>
<tbody>
<tr>
<td>Vasa (rabbit)</td>
<td>1:5000</td>
<td>Ruth Lehmann</td>
</tr>
<tr>
<td>Bam (mouse)</td>
<td>1:500</td>
<td>DSHB bam</td>
</tr>
</tbody>
</table>

**RNA-seq**

Ovaries were dissected for each sample in cold Dissection buffer (PBS 1X, 0.01% Tween 20, 100 µg/ml cycloheximide) and immediately stored at -80°C after dissection. Frozen samples were homogenised in Polysome extraction buffer (50 mM Tris pH 7.5, 5 mM MgCl2, 150 mM NaCl, 0.5% Triton x-100, 1 mM DTT, 100 µg/ml cycloheximide, protease inhibitor, 25 u/ml Turbo DNase) using an electrical pestle and then further disrupted by passing 20 times through a 26-gauge needle. The lysate was centrifuged at 20,000g for 10 min at 4°C to pellet tissue debris. Total RNA was isolated by hot phenol-chloroform extraction and quantified using Qubit RNA High Sensitivity Assay Kit (Invitrogen). 1.5 µg of total RNA was used for rRNA depletion by binding of complementary oligos and treatment with RNase H. Libraries were generated using NEBNext Ultra Directional RNA Library Prep Kit for Illumina according to manufacturer’s instructions. Libraries were multiplexed using the NEBNext Multiplex Oligos for Illumina and sequenced in paired-end 150 nt-long reads on an Illumina NovaSeq 6000.
Ribo-seq
250 pairs of ovaries were dissected for each sample in cold Dissection buffer (PBS 1X, 0.01% Tween 20, 100 µg/ml cycloheximide) and stored at -80°C. As for RNA-seq, frozen samples were homogenised with an electrical pestle in Polysome extraction buffer (50 mM Tris pH 7.5, 5 mM MgCl2, 150 mM NaCl, 0.5% Triton x-100, 1 mM DTT, 100 µg/ml cycloheximide, protease inhibitor, 25 u/ml Turbo DNAsel), then passed 20 times through a 26-gauge needle. The lysate was centrifuged at 20,000g for 10 min at 4°C to pellet tissue debris. 100-140 µg of total RNA was used for Ribo-seq. Total RNA was treated with 1.25 units of RNase I (Ambion) per µg RNA, and the quenched with SuperaseIn (0.8 units per µg RNA). The sample was brought to 1 ml by adding Polysome extraction buffer and subjected to a 3 ml 34% sucrose cushion, by centrifugation at 70,000 rpm for 5.5 hr at 4°C. The resulting pellet was resuspended in nuclease free water with 1% SDS. RNA was extracted with hot phenol-chloroform and purified by Zymo RNA Clean & Concentrator kit. rRNA was depleted as for the RNA-seq through binding of complementary oligos and treatment with RNase H71,72. Size selection was performed on a 15% Urea PAGE Gel, cutting the band corresponding to 28 to 34 bp. RNA was purified by ZYMO small-RNA PAGE recovery kit. RNA was treated with 1 µl of T4 PNK for end repair and libraries were produced with the NEBNext Small RNA Library Kit, according to manufacturer’s instructions. Libraries were size selected at 148-154 nucleotides, corresponding to ligated constructs from 28 to 34 nt RNA fragments. Libraries were multiplexed using the NEBNext® Multiplex Oligos for Illumina® and sequenced in single-end 50 nt-long reads on an Illumina HiSeq 4000.

Data analysis
Trim galore, integrating the trimmer tool cutadapt73, was used for adapter trimming and quality control of both (paired-end) RNA-seq and (single-end) Ribo-seq data. Subsequently, random 4 nucleotides (4N adapters) from both 3′- and 5′-ends, introduced during library preparation, were removed from Ribo-seq reads after excluding redundant reads. Trimmed reads were aligned to non-coding RNA reference sequences (flybase, dmel_r6.39) using Bowtie74 for Ribo-seq and Bowtie275 for RNA-seq data and then non-matching reads were mapped to the Drosophila melanogaster reference genome dm6 using STAR76. For quality control and benchmarking of Ribo-seq data we used the R package ribosomeProfilingQC (Ou and Hoye 2022). Transcript abundance was quantified and differentially expressed genes were identified using Cuffdiff v2.2.177. For analysis of translation efficiency, read counts were generated using featureCounts78. Differences in translation efficiency and classification of gene regulation was analysed using deltaTE61. All analyses were performed with two samples, each with two biological replicates. GO analysis was performed using FlyMine79. PCA analysis was performed using the precomp function in R, using scaled variables.
Supplementary Figures

(A) | RNA-seq | mRNA | Hoechst | phalloidin | mRNA |
--- | --- | --- | --- | --- | ---

Figures S1 – RNA-seq in synchronised differentiating GSCs recapitulates changes in gene expression during normal differentiation (Related to Figure 2)

(A) As in Figure 2A, smFISH in wild type germaria (right) was used to validate the RNA-seq results (left). Staining for DNA (Hoechst, blue), actin (phalloidin, gray), fusome (α-spectrin, yellow), and mRNA transcript of interest (smFISH, red and grayscale): (i) cuff, (ii) thymidylate synthase, (iii) CG11674, (iv) CG32814 (eggplant), (v) CG14545, (vi) blanks. Scale bars are 15 µm. (B) The 0.5 hr AHS sample was used to exclude 67 genes which showed a significant change of >3-fold between the 0.5 AHS and the no HS samples (with expression >10 FPKM in one of these samples). The expression of these 67 genes is shown as log2foldchange at each time point relative to no HS. (C) GO analysis of the excluded 67 genes shown in (B) found a significant enrichment in terms associated with a heat shock response. Colour of the bar indicates p value of the enrichment.

![Figure S1](image-url)
**Figure S2** – RNA-seq reveals changes in mRNA level during GSC differentiation (Related to Figure 2)
(A) PCA analysis of the RNA-seq in the differentiating GSC time course. (B) Gene ontology analysis of genes upregulated or downregulated in the RNA-seq at 9 hr AHS or 38 hr AHS compared to ‘no HS’. Colour of the bar indicates p value of the enrichment, grey = p > 0.05 (C) RNA-seq FPKM for genes in two gene ontology groups: mitotic DNA replication (i) and P granule (ii). Genes with a >1.6-fold change in gene expression are highlighted.
Figure S3 – RNA-seq in synchronised differentiating GSCs recapitulates changes observed in scRNA-seq from whole ovaries (Related to Figure 2)

Heatmaps illustrating RNA-seq expression level across the time course. Each row represents one gene and expression level is scaled per gene such that black represents the mean expression across the time course, gold represents a 25% higher expression than the mean, and cyan represents a 25% lower expression than the mean. Gene groups shown are those identified by scRNA-seq pseudotime analysis by Rust et al., 2020 (illustrated in Fig 2m in that paper): (A) marker genes for each time point from GSC to 16cc, (B) Spindle genes which were upregulated during the mitotic stages of differentiation, (C) synaptonemal complex genes were upregulated around the 8cc stage, and additional candidate synaptonemal complex genes with a similar RNA expression pattern.
Figure S4 – Ribo-seq quality control (Related to Figure 3)

(A) $R^2$ show good correlation between Ribo-seq replicate samples at each time point. (B) Distribution of read length in nucleotides for Ribo-seq libraries at each time point. (C-F) Ribo-seq quality control for the ‘no HS’ sample only for illustration. (C) 90.73% of reads mapped to the CDS on the sense strand of genes compared to 0.05% mapping to the CDS on the antisense strand. (D) 8.87% of reads map to the 5’ UTR, 2.20% map to the 3’ UTR, and mapping to introns and intergenic regions is negligible. (E) P site mapping shows a strong three nucleotide periodicity, with highest frequency at the start codon. (F) Metagene analysis plot showing read distribution in 5’ UTR, CDS and 3’ UTR regions, shows consistent coverage across the CDS with the expected bias at the start codon.
Figure S5 – Ribo-seq reveals changes in translation during GSC differentiation (Related to Figure 4)

(A) Changing expression of different genes in the RNA-seq (solid line) or Ribo-seq (dashed line). Ribo-seq in orb and bgcn reflects the RNA-seq, but for dhd and wit there is additional regulation at the level of translational efficiency. (B) PCA analysis of the Ribo-seq in the differentiating GSC time course. (C) Gene ontology analysis of genes upregulated or downregulated in the Ribo-seq at 9 hr AHS or 38 hr AHS compared to ‘no HS’. Colour of the bar indicates p value of the enrichment, grey = p > 0.05. (D) Ribo-seq FPKM for genes in three gene ontology groups: mitotic DNA replication (i), P granule (ii) and cytosolic ribosome (iii). Genes with a >1.6-fold change in gene expression are highlighted (except in iii due to too many genes).
**Figure S6 – Different groups of genes are regulated by mRNA level or TE (Related to Figure 5)**

Gene ontology enrichment analysis of genes with a >1.6-fold change in Ribo-seq at 9 hr AHS or 38 hr AHS compared to ‘no HS’. Enrichment analysis was performed separately for genes which are upregulated or downregulated, as well as for those regulated primarily by mRNA level (>1.6-fold change in RNA-seq) or by TE (>1.6-fold change in TE).

**Supplementary Table 1.** RNA-seq FPKMs for genes with a heat shock effect over 3-fold (significant, and over 10 FPKM), and list of contaminants.

**Supplementary Table 2.** rRNA depleted RNA-seq FPKMs for each time point, after filtering (>10 FPKM, removal of contaminants)

**Supplementary Table 3.** Ribo-seq FPKMs for each time point, for the genes as in Supplementary Table 2