Supplemental Data

*"***groHMM: A Computational Tool for Identifying Unannotated and Cell Type-Specific Transcription Units from Global Run-On Sequencing Data"**

Chae *et al.* **(2015)**

1) Supplemental Tables

Table S1. Public human GRO-seq data sets mined using groHMM.

Table S2. List of HMM and non-HMM based broad peak callers and their applicability to the analysis of GRO-seq data.

Table S4. Comparison of transcription units called by groHMM using optimal parameters versus the average of all 50 explored parameter sets for *D. melanogaster* **GRO-seq data.**

Table S6. Top ten GO terms for the cell type-specific enhancer clusters.

Table S7. Top ten GO terms for the non-cell type-specific enhancer cluster.

1) Supplemental Figures

Figure S1. Parametric space for explored 100 models comparing three transcript callers: groHMM, SICER, and HOMER.

(A) Number of transcripts called.

(B) Median transcript length.

(C) Total error rate for two types of error ('merged annotation error' and 'dissociated annotation error').

(D) Number of occurrences of 'merged annotation error.'

(E) Number of occurrences of 'dissociated annotation error.'

Figure S2. Variations in TUA with gene expression patterns.

(A) Evaluation of TUA when varying EDR (i.e., the smoothness of expression patterns) for mRNA genes.

(B) Evaluation of TUA when varying EDR (i.e., the smoothness of expression patterns) for lncRNA genes.

(C) TUA of called transcripts for well-expressed lncRNA annotations (n = 2,403). Ten percent of the annotations were bootstrapped with replacement ($n = 100$).

Figure S3. Functional analysis of cell type-specific enhancers in three cell types. *[See following page for the figure]*

In order to infer the function of the cell type-specific enhancers that we identified above, we used Gene Set Enrichment Analysis (GSEA) [39].

(A) Association matrix for the cell type-specific enhancers with functional gene sets. We determined the correlation of the transcription of each protein-coding gene with the transcription of each of the 1,052 cell type-specific enhancers. We then ranked the protein-coding genes based on the strength of their correlations and used these rankings to assign enrichment scores for all GSEA categories (i.e., gene ontology, or GO, terms) for each enhancer. Next, we performed hierarchical clustering analysis, displaying the normalized GSEA enrichment scores for each enhancer. Each row is a GO term with its associated normalized GSEA enrichment scores and each column represents an enhancer. This analysis identified seven clusters (**Table S5**). The GO terms in the clusters represent the characteristics of the cell type in which the enhancers are active. Red = positive association; green = negative association. The clusters were identified by using the cuttree function of R.

(B) Summary heatmap for the clusters shown in (A). The median values of each cluster were used for a more simple visual representation.

(C) Association matrix of non-cell type-specific enhancers with functional gene sets, as in (A). Analysis of 837 non-cell type-specific enhancers yielded fewer clusters and failed to group the enhancers from each cell type (**Table S6**).

(D) The top ten GO terms for cluster $4(n = 57)$ summarized by REVIGO. The p-values for all terms were < 0.0001 and ordered by 'dispensability,' which represents the non-redundancy of the term in the group [57]. The cluster 4 enhancers (panel A), which are active in LNCaP cells treated with dihydrotestosterone, are associated with GO terms related to steroid signaling, endocrine processes, and cellular signaling.

(E) Top ten GO terms for cluster 1 ($n = 124$) in (D) summarized by REVIGO (p-values ≤ 0.001 and ordered by dispensability). The cluster 1 enhancers from the non-cell type-specific enhancer analysis (panel C), which are active in multiple cell types, are associated with GO terms related to a broader array of cellular processes.

(F) Pie charts showing regulation of enhancer transcription by treatment in each cell type (MCF-7, estradiol; LNCaP, dihydrotestosterone; AC16, tumor necrosis factor alpha). Regulation was called using FDR \leq 1% for MCF-7 and AC16, and p-value \leq 0.001 for LNCaP with edgeR. The data from the MCF-7, LNCaP, and AC16 cells gave us a unique opportunity to address this questions, given the availability of GRO-seq data sets from hormone-treated cells (MCF-7, estradiol; LNCaP, dihydrotestosterone; AC16, tumor necrosis factor alpha). When compared to the basal (untreated) condition, the treatments affected (either upregulated or downregulated) the transcription of between 25% and 65% of the cell type-specific enhancers within a given cell type, with the effects of the estradiol treatment in MCF-7 cells being most pronounced.

(G) Effects of treatments on mRNA and lncRNA transcript expression in three cell types. Pie charts showing the percent of protein-coding transcripts *(left)* and lncRNA transcripts *(right)* regulated by treatment in each cell type (MCF-7, estradiol; LNCaP, dihydrotestosterone; AC16, tumor necrosis factor alpha). Regulation was called using FDR < 1% for MCF-7 and AC16, and p-value < 0.001 for LNCaP using edgeR. The proportions of regulated protein-coding (mRNA) or lncRNA transcripts were similar to the proportions of regulated enhancer transcripts for each cell types.

Figure S4.

