# PEACE: Parallel Environment for Assembly and Clustering of Gene Expression

D.M. Rao<sup>1</sup>, J.C. Moler<sup>1</sup>, M. Ozden<sup>1</sup>, Y. Zhang<sup>1</sup>, C. Liang<sup>1,2\*</sup> and J.E. Karro<sup>1,3\*</sup>

- <sup>1</sup> Department of Computer Science and Software Engineering,
- <sup>2</sup> Department of Botany,
- <sup>3</sup> and Department of Microbiology, Miami University, Oxford, Ohio, USA

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#### **ABSTRACT**

We present PEACE, a tool for the high-throughput clustering of transcript fragment sequences produced Next Generation or Sanger Sequencing technologies. Installed and easily run through an easily downloaded GUI, PEACE can process large data sets, grouping the fragments by gene association, faster and with greater sensitivity than competing clustering tools such as WCD, and with considerably greater sensitivity than the implicit clustering preformed by assembly tools such as CAP3. Through the GUI the user can collect cluster statistics and examine specific clusters for more comprehensive study. And by using PEACE as a "pre-assembly" tool to allow the application of an assembly tool to the resultant individual clusters, the user gains both a significant increase sensitivity of the clustering method and a large decrease in overall runtime for the clustering and assembly together.

### INTRODUCTION

Understanding an organism's transcriptome, the set of (spliced) transcripts expressed by genes of the organism, is a vital step in understanding the full functional and organization roll of the genome in the life cycle of any organism. Studying the transcriptome has led to gene discovery, provided information of protein isoforms, and generally helped shed ight on the biological processes both controlling and dictated by the genome (1). But in order to get access those transcripts, we must first deal with the fragmented data produced by both Next Generation and traditional Sanger sequencing technology.

Until recently, a primary method of gaining access to a transcriptome sequence was through the use of Express Sequence Tags (ESTs), which were primarily single-pass cDNA sequences derived from transcribed mRNAs and traditionally sequenced by Sanger Sequencing technology. More recently, Next-Generation Sequence (NGS) technologies have replaced the Sanger Sequencing techniques, allowing for considerably greater depth and coverage of the transcriptome. For example, ESTs in GenBank dbEST are increasingly the product of NGS technologies such as 454 pyrosequencing, which enables the sequencing of novel and

rate transcripts at a considerably higher rate of coverage than did Sanger Sequencing (2, 3). But from a computational perspective, this is a mixed blessing: while NGS provides immense quantities of new information, it does provides immensely larger data sets for analysis, requiring even more efficient algorithms than were already needed.

Given a set of transcript fragments, produced by either sequencing technology, sampled from multiple transcripts across the genome, a necessary first step is that of clustering: separating the fragments according to the transcript from which they were derived. While clustering is frequently preformed implicitly by assembly tools such as CAP3 (4), clustering the data as a "pre-assembly" step has a number of advantages. Most significant among these: preforming this step will allow the application of the assembly tool to individual clusters – saving significant amounts of time due to the reduced input size (5). But clustering is a computationally challenging problem, especially when done ab initio instead of relying on access to a sequenced genome. Even with the number of ESTs produced using Sanger sequencing, the runtime and memory requirements to cluster based on pairwise sequence alignments make such an approach infeasible. The much larger data-set sizes produced by the nextgeneration sequencing technologies exacerbates this problem.

Here we present PEACE, a tool for the ab initio clustering of transcript fragments by gene, applicable to both short-read and traditional Sanger Sequencing technologies. Available through the www.peace-tools.org website, the PEACE GUI allows the user to both easily install (locally or remotely) and run the computational tool, as well as providing various tools for result analysis. PEACE can be run in single or multi-processor mode, and requires no Internet connectivity past the initial download. Using a minimum spanning tree (MST) based clustering strategy (6, 7) and the  $d^2$  sequence distance function (8), PEACE generates the MSTs using a tailored version of Prim's algorithm (9). The result is an easyto-use clustering tool that is considerably more sensitive to sequencing errors than the competing WCD tool (5) and others in the literature without any sacrifice of performance (e.g. (10), (11), (4), (12), (13), (14), (15), (16)).

<sup>\*</sup>to whom correspondence should be addressed

#### PEACE: INSTILLATION AND USE

The PEACE GUI, available from the PEACE website (www.peace-tools.org), handles the details of both instillation and job management and provides tools for the initial analysis of the results. By downloading and executing the peace. jar GUI Java program on a Linux, Windows or OS-X machine running Java 1.6+, the user is presented with the PEACE management tool. Through this the user can install the computational tool and preform a clustering of a data file, view an initial analysis of the clusters and produce files containing subsets of the clusters for assembly by a tool such as CAP3 (4). A typical (first) use of the tool would generally require the the following steps.

Tool Instillation: Installing the GUI requires only that the Java file be downloaded from the website. To install the PEACE tool onto a local or remote machine, the user will select from within the GUI the Server > Add New Server tab from the GUI main menu (Figure 1(a)), which then starts the install wizard that will prompt for the appropriate information. Figure 1(b) illustrates the request for server information; in this example, the user has chosen to install the PEACE computational tool on a remote machine and is providing the necessary connection information. Server information is persistent between GUI sessions, giving the user access to PEACE on the target machine as needed.

Job Processing: After importing the target file into the GUI (File > New Data Set), the user can start a new job with the Job > Job to Complete Clusters tab and follow the wizard menus. Figure 1(c) illustrates the process of specifying the number of processors available (if running on a machine supporting the OpenMPI protocol - which will be determined during job instillation). Once executed, the GUI will manage the job thread, alert the user when the job is completed (or when the user next runs the GUI after completion) and copy the final results back to the local machine if necessary.

Result Analysis: Once the resulting clusters have been computed and (if necessary) copied back, the user has several options for analysis:

- Export: The user can export the contents of one or more clusters into a fasta file, obtaining a subset of the original file containing the sequences corresponding to the selected clusters ready for processing by an assembly tool (e.g. CAP3 (4)).
- View Clustering: The user may view a list of all clusters, expanding selected clusters to a list of all fragments (illustrated in Figure 1(d)).
- Classified Summary Graph: The user may view a distribution of cluster sizes in a linear or log scale. Further, the user may set up a classifier, associating certain patterns with specific colors. These patterns will be matched against the fragment header information from the original fasta file, allowing the overlay of a colored cluster size distribution does clusters containing matching elements in a specified colors. For example, if the sequence names contain unique string patterns denoting different cDNA libraries, the classifier can help the user to determine the differential expression

profiles of different libraries for a given cluster by color coding the strains. The method of setting up a these classifiers, and the resulting histogram, is illustrated in Figure 1(e).

Extensive documentation for the tool has been posted on the PEACE website, as well as links to several videos demonstrating PEACE use and capabilities.

## **METHODS**

The clustering performed by PEACE is based on the use of minimum spanning trees (MSTs), known to be an effective approach for narrow band single linkage clustering (6, 7). We use the  $d^2$  distance measure (8) to represent edge weights. Prim's algorithm (9) then allows us to efficiently calculate an MST from which we can infer a high-quality clustering solution.

The  $d^2$  distance measure used to assign edge weights is an alignment-free measurement of sequence distance that can be calculated significantly faster than a Smith-Waterman alignment (8, 17).  $d^2$  works by comparing the frequency of words (strings of a fixed length) appearing in a limited region of each string. Fragments overlapping by a sufficient length will share neighborhoods of enough similarity to ensure a small distance even in the presence of a moderate number of base errors. Our method of calculating  $d^2$ , the two-pass  $d^2$ heuristic, is a refinement of that used in the WCD clustering tool (5). We augment the computation with a preliminary pass that narrows down the two regions of maximum similarity, and then explore those local regions in more detail.

We can model the fragment input as a weighted, undirected graph: the fragments are represented as nodes, with  $d^2$ sequence distances assigned to the connecting edges as weights. Conceptually, we want to remove each edge exceeding a threshold score from the complete graph, and define our partitions by the remaining connected components of the graph. An edge with a large weight connects fragments which are likely unrelated; once such edges are removed, the components define a series of overlaps. Hence those fragments that can still be connected by some path correspond to the same gene. However, such an approach requires the calculation of all edge weights. Such a task is infeasible both in terms of runtime and memory usage for the data set sizes we expect to process.

We approach the problem by generating a minimum spanning tree of the described graph, then removing edges exceeding our threshold. By using Prim's algorithm we are able to calculate edge weights on-the-fly (reducing memory requirements), and we can skip the calculation of a majority of edge distances using the u/v and t/v filtering heuristics employed in WCD (5). These heuristics allow us to quickly dismiss many of the edges as too large without the need to apply the full  $d^2$  algorithm (see Sections A and B of the Supplementary Materials for more details).

## **RESULTS**

We have tested PEACE on both simulated and real data, comparing results against those produced by the WCD clustering tool (5) and the CAP3 assembly tool (4) (which calculates a cluster as an intermediate step). For our simulation tests we used the **ESTSim** tool (18) to generate simulated transcript fragments under varying models of error (Supplementary Materials, Section), generating the fragments from the list of 100 zebra fish genes used in the WCD testing (5). Tool parameters were taken to match, as closely as possible, those used in the WCD study (see supplementary materials). Our primary methods of quality assessment were sensitivity (the fraction of fragment pairs from the same gene that were correctly clustered together) and Type 1 error (the fraction of genes that were divided between clusters) (5, 19); other measurements are also discussed in the supplementary materials. In Figure 2(left) we see a significant improvement of PEACE (blue) over WCD (green) and CAP3 (black) in its sensitivity to sequencing errors, while in Figure 2(right) we see a comparable improvement in Type 1 error. In Figure 3 we look at the number of singleton clusters (fragments not joined to any cluster), which will not occur in our simulated sets; we again see significant improvement in PEACE. The sequential runtime of PEACE is slightly improved over that of WCD (see Supplementary Materials, Figure S4).

In applying the tools to real data, we started with a set of approximately 190,000 clean Sanger ESTs derived from the *Chlamydomonas reinhardtii* genome (20). To compute sensitivity we used the *gmap* tool (21) to map the fragment set to the genome, taking this as our reference clustering. We see slight improvements in PEACE over WCD in both sensitivity and Type 1 error (both significantly out preforming CAP3). Using the mouse fragment set used in Hazelhurst *et al.* (5) PEACE again shows slight improvements in sensitivity and Type 1 error rates, with an 18% speed up for PEACE (see Supplementary Materials, Section, for more details).

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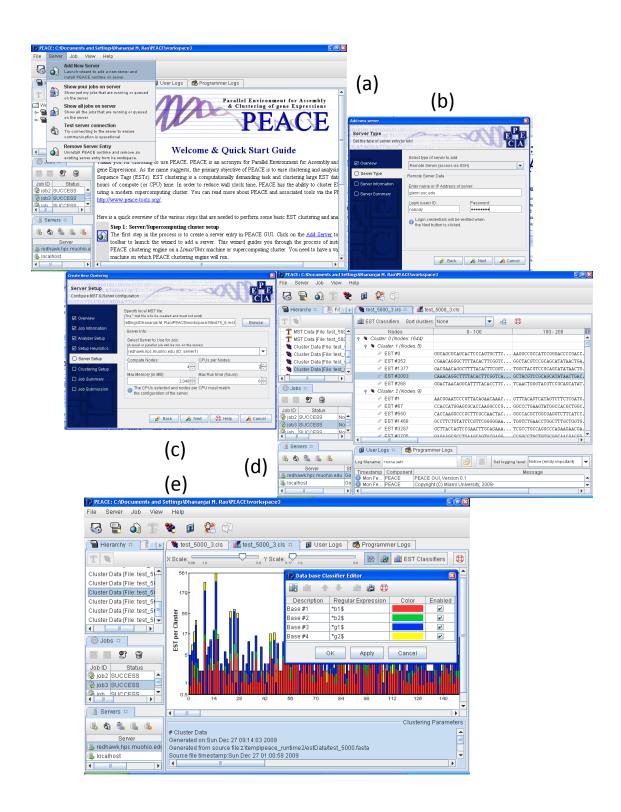
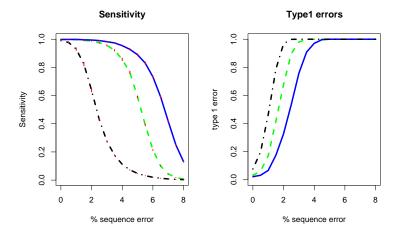
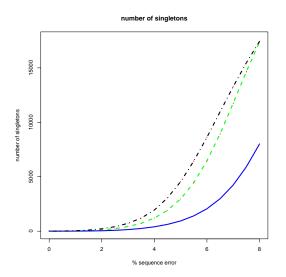


Figure 1. Screenshots of the PEACE GUI during execution, including (a) GUI Welcome and server instillation menu; (b) setup wizard for installing the computational tool on a remote server; (c) execution wizard for starting a selected job to be executed in parallel mode; (d) basic cluster output; and (e) histogram view of cluster results and classifier editor.



**Figure 2.** Comparisons of Sensitivity and Type 1 error, based on the average over 30 Simulated fragment sets derived from 100 zebra fish genes (see Supplementary Materials, Section, for more details). Blue/Solid = PEACE, Green/Dash = WCD, Black/Dot-Dash = CAP3; vertical tics = 95% confidence intervals on estimates. Intervals are not presented for Type 1 error due to the extremely small variance.



**Figure 3.** Average number of fragments flagged as singletons by each tool when run on the simulated sequences; correct answer in all cases is zero. Blue/Solid = PEACE, Green/Dash = WCD, Black/Dot-Dash = CAP3.