Using RAMPAGE to identify and annotate promoters in insect genomes

R. Taylor Raborn $^{\star 1}$ and Volker P. Brendel 1,2

¹Department of Biology, Indiana University ²School of Informatics and Computing, Indiana University

Department of Biology Indiana University 212 S. Hawthorne Drive 205 Simon Hall, Bloomington, IN 47401, USA http://www.brendelgroup.org

Abstract. Application of Transcription Start Site (TSS) profiling technologies, coupled with large-scale next-generation sequencing (NGS) has yielded valuable insights into the location, structure and activity of promoters across diverse metazoan model systems. In insects, TSS profiling has been used to characterize the promoter architecture of *Drosophila melanogaster* [1] and subsequently was employed to reveal widespread transposon-driven alternative promoter usage in the fruit fly [2]. In this chapter we discuss the computational analysis of the experimental data derived from of one TSS profiling method, RAMPAGE (RNA Annotation and Mapping of Promoters for Analysis of Gene Expression), that can be used for the precise, quantitative identification of promoters in insect genomes. We demonstrate this using the software tools GoRAMPAGE [3] and TSRchitect [4], providing detailed instructions with the aim of taking the user from raw reads to processed results.

Keywords: cis-regulatory regions, promoter architecture, transcription initiation, transcription start sites (TSSs)

1 1 Introduction

2 1.1 TSS Profiling Identifies Promoters at Genome-Scale

- 3 The promoter, which is defined in eukaryotes as the genomic region bound by
- 4 RNA Polymerase II immediately prior to transcription initiation [5], is the pri-
- 5 mary locus of the regulation of gene expression. The identification of promoter
- 6 regions is necessary for understanding the cis-regulatory signals controlling gene
- 7 expression in an organism, and is also important for genome annotation. How-
- ever, despite the rapid accumulation of genome sequences across metazoan and
- arthropod diversity, accurate annotation of promoter regions remains sparse.
- This is because—absent empirically-defined information—precisely identifying

^{*} Correspondence: rtraborn@indiana.edu

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sequence motifs that demarcate the promoter is unreliable. In contrast with current *in silico* approaches, direct mapping of TSSs identifies the location of the core promoter. Cap Analysis of Gene Expression (CAGE) [6], one of the first methods devised to identify 5'-ends of mRNAs at large-scale, involves selective capture of 5'-capped transcripts, first-strand reverse-transcription and ligation of a short oligonucleotide (CAGE tag).

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CAGE was initially utilized by the FANTOM (Functional Annotation of the Mammalian Genome) consortium to identify promoter architecture in human and mouse [7], providing the first glimpse of the global landscape of transcription initiation. At the onset of the next-generation sequencing (NGS) era, CAGE was coupled with massively-parallel sequencing to define 5'-mRNA ends at large scale. This advance provided more extensive coverage of the expressed transcriptome and provided increased sensitivity for quantitative measurements of promoter activity.

26 1.2 Promoter Architecture of Drosophila melanogaster

Hoskins and colleagues [1] performed CAGE in D. melanogaster as part of the 27 modENCODE consortium, identifying promoters at large-scale and characteriz-28 ing the promoter architecture of an insect genome for the first time. The authors 29 found that TSS distributions at *Drosophila* promoters exhibit a range of shapes that can be generally grouped into two major classes: peaked and broad. This 31 confirmed the original finding of Rach and colleagues [8], which was done us-32 ing publicly-available expressed sequence tags (ESTs). Peaked promoters have a 33 single, major TSS position occupying a narrow genomic region, whereas broad promoters lack a single, major TSS and contain TSSs across a wider region [8, 9]. 35 The authors also showed a strong association between promoter class and motif 36 composition (consistent with previous findings [8, 10]). Peaked promoters were associated with positionally-enriched cis-regulatory motifs including TATA, Initiator (Inr) and DPE (Downstream Promoter Element), while broad promoters 39 contained an enrichment of less-well characterized motifs, including Ohler6 and 40 Ohler [11]. The existence of at least two promoter classes appears to be con-41 served among metazoans and has been reported (using TSS profiling methods) 42 in insects, cladocerans [12], fish [13] and mammals [14, 9]. 43

44 1.3 Promoter Structure of Insects

Beyond *D. melanogaster*, few investigations have utilized TSS profiling in insect genomes. As a consequence, what is known about promoter architecture in insects is largely restricted to the *Drosophila* genus. As part of the modENCODE effort, CAGE was performed in multiple tissues and developmental stages of the *Drosophila pseudoobscura*. TSSs were found to be highly similar between species: 81% of TSSs of aligned, CAGE-identified TSSs from *D. pseudoobscura* were positioned within 20nt of their counterparts in *D. melanogaster*. An enrichment of

the CA dinucleotide was detected at the TSS ([-1, +1]), and the motifs corresponding to TATA, Inr and DPE were positioned at the same locations relative to the TSS in both species.

The only other insect species for which TSS profiling has been applied is the 56 Tsetse fly (Glossina morsitans morsitans) [15]. Using TSS-seq (specifically Oligocapping; for details see [16]), the authors identified 3134 promoters associated 58 with 1424 genes. The authors found a preference for CA and AA dinucleotides at 59 the TSSs and observe the major core promoter elements observed in *Drosophila*: 60 TATA, Inr, DPE, in addition to MTE (Motif Ten Element). As in D. melanogaster, 61 peaked promoters were more likely to contain TATA and Inr than broad promot-62 ers. While the taxonomic sampling of species for TSS profiling has been limited, 63 the existing studies are sufficient to provide a general picture of insect promoter architecture. A major demarcation between the promoter architecture of insects and mammals appears to be the large fraction of mammalian promoters found 66 in CpG islands [15]. CpG island promoters (CPIs) form the largest class of pro-67 moter in mammals [17]; by contrast, CPIs are not known to exist as a class in invertebrates.

70 1.4 Paired-end TSS Profiling with RAMPAGE

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The most recent major methodological advance in TSS Profiling is RAMPAGE 71 [2, 18], a protocol for 5'-cDNA sequencing that combines cap trapping and template-72 switching with paired-end sequence information. A key advantage of generating 73 paired-end sequence is transcript connectivity, which provides a direct link be-74 tween a given 5'-end and its associated mRNA molecule [2]. Because short or 75 spurious RNAs are found within the transcriptome, transcript connectivity allows the TSSs (and thus promoters) of full-length mRNAs to be unambiguously 77 identified, which benefits genome annotation and improves interpretation of tran-78 script species. 79

stages across the life cycle of *D. melanogaster*, generating a comprehensive gene expression and promoter atlas for fruit fly and demonstrating the utility of RAM-PAGE. RAMPAGE is currently being applied as part of the latest iteration of ENCODE to identify promoters in human, but as of this writing it has not been applied to any non-*Drosophila* insect model system. In anticipation of the future application of TSS profiling into other insect model systems, we discuss in this chapter a well-documented protocol for the computational processing and

Batut and colleagues [2] generated libraries from total RNA isolated from 36

analysis of RAMPAGE data, using selected libraries from Batut *et al.* [2]. This method consists of two parts: first, we discuss how to process, filter and align the sequenced RAMPAGE libraries to the *D. melanogaster* genome. Second, we

show how to identify TSSs and promoters from the aligned sequences and associate them with coding regions. In closing, we will consider further applications

of this data and discuss the utility of reproducible workflows in bioinformatic analysis.

96 2 Materials

The analyses described herein require a workstation capable of doing modern 97 bioinformatics; minimally a reasonably-appointed laptop. An intermediate un-98 derstanding of the Linux/Unix command line will be extremely useful, although 99 we make efforts to explain the procedures with clarity. In addition, it will likely 100 be necessary for the participant to have superuser privileges on the machine. 101 If you do not have a machine (or have access to one) that meets these re-102 quirements, it is recommended that you consider cloud-based cyberinfrastructure, including Amazon Web Services (AWS; https://aws.amazon.com/), Cy-104 Verse (http://www.cyverse.org/) [19], or JetStream (https://jetstream-cloud.org/) 105 [20]. The former is a well-known pay-per-use solution, while the latter two are 106 NSF-funded resources that makes compute allocations freely available to the 107 public. 108

109 2.1 Hardware

- 1. x86-64 compatible processors
- 2. At least 8GB RAM
- 3. 30GB+ hard disk space

113 2.2 Operating System

- 64 bit Linux (preferred) or Mac OS X (with Command Line Tools from
 XCode)

116 2.3 Software

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Below is a list of the software packages required for this demonstration (see Note 1).

120 Sequence retrieval

1. SRA Toolkit [21] (https://www.ncbi.nlm.nih.gov/sra/docs/toolkitsoft/)

122 GoRAMPAGE

- 1. GoRAMPAGE [3] (https://github.com/brendelGroup/GoRAMPAGE)
- 2. fastq-multx [22] (https://github.com/brwnj/fastq-multx)
- 3. FASTX-Toolkit [23] (http://hannonlab.cshl.edu/fastx_toolkit/Index.html)
- 4. TagDust2 [24] (https://sourceforge.net/projects/tagdust/)
- 5. Samtools [25] (http://www.htslib.org/doc/samtools.html)
- 6. STAR [26] (https://github.com/alexdobin/STAR)

129 TSRchitect

- 130 1. R (v. 3.4 and up) [27] (https://www.r-project.org/)
- 2. Bioconductor (v. 3.5 and up) [28] (http://bioconductor.org/)
- 3. TSRchitect [4] (http://bioconductor.org/packages/release/bioc/html/TSRchitect.html)
- 4. Various R package dependencies (see Methods)

134 2.4 Demonstration

```
We created an online demonstration (demo) to serve as a companion to this chapter, which contains both scripts and select files to assist you in completing this tutorial. Please find the repository at https://github.com/brendelgroup/GoRAMPAGE/demo/MMB (see Note 2).

2.5 Installation of R packages

For installation of the software listed above, please follow the instructions provided by each respective package. Part of our analysis will require the use of R packages found in the Bioconductor suite [28]. To install Bioconductor, please
```

```
type the following from an R console:
143
    source("https://bioconductor.org/biocLite.R")
    biocLite()
    We will use the R package TSRchitect to identify promoters from aligned RAM-
146
    PAGE libraries. Prior to running the analysis, it will be necessary to install a
147
    series of prerequisite packages to TSRchitect from Bioconductor. Please install
    these packages, followed by TSRchitect (as before, from an R console):
    source("https://bioconductor.org/biocLite.R")
    biocLite(c("AnnotationHub", "BiocGenerics", "BiocParallel",
151
     "ENCODExplorer", "GenomicAlignments", "GenomeInfoDb",
152
     "GenomicRanges", "IRanges", "methods",
153
     "Rsamtools", "rtracklayer", "S4Vectors",
     "SummarizedExperiment"))
155
    biocLite("TSRchitect")
```

Finally, please confirm that TSRchitect has been installed correctly by loading

161 3 Methods

it from your R console as follows:

library(TSRchitect) #loading TSRchitect

158

162 3.1 Retrieving the RAMPAGE sequence data from NCBI

To begin our analysis, we must download the RAMPAGE data to our workstation. We will utilize tools provided by the SRA Toolkit, which should already be installed on your machine (see Materials). The command fastq-dump allows one to directly retrieve data from the GEO database using the appropriate identifier(s). While there are 36 RAMPAGE libraries in the Batut et al. paper, we will select a subset of these to analyze here. We will compare samples from selected embryonic (E01h-E03h) and larval (L1-L3) tissues, representing the beginning and end of embryonic development. For more information about

```
the experiment and the available RAMPAGE libraries, please see the following
171
    link: https://www.ncbi.nlm.nih.gov/Traces/study/?acc=SRP011193.
172
   First, let's proceed with downloading the libraries from early embryonic tissues
174
    (see See Note 3). We will make a new folder (entitled "fastq_files/") to
175
   house these files.
    mkdir fastq_files
    cd fastq_files
178
    fastq-dump --split-files SRR424683
    fastq-dump --split-files SRR424684
181
    fastq-dump --split-files SRR424685
182
    We continue by downloading the data from late larval tissues.
    fastq-dump --split-files SRR424707
    fastq-dump --split-files SRR424708
    fastq-dump --split-files SRR424709
186
    Once the download of the aforementioned files are complete, you should see a
187
    total of 12 (6 x 2) separate fastq files in your current working directory:
   ls -l *.fastq | wc -l
    3.2 Creating symlinks to the files
    Our workflow expects fastq files that have the format "*.R1/R2.clipped.fq".
   Rather than rename them, we can simply create brand new symbolic links (sym-
192
   links) to the files, as follows:
193
    cd .
   mkdir -p output/reads/clipped
195
    cd output/reads/clipped
196
197
   #embryonic libraries
   ln -s ../../fastq-files/SRR424683_1.fastq E01h.R1.clipped.fq
199
   ln -s ../../fastq-files/SRR424683_2.fastq E01h.R2.clipped.fq
   ln -s ../../fastq-files/SRR424684_1.fastq E02h.R1.clipped.fq
   ln -s ../../fastq-files/SRR424684_2.fastq E02h.R2.clipped.fq
   ln -s ../../../fastq-files/SRR424685_1.fastq EO3h.R1.clipped.fq
203
   ln -s ../../fastq-files/SRR424685_2.fastq E03h.R2.clipped.fq
204
   #larval libraries
   ln -s ../../fastq-files/SRR424707_1.fastq L1.R1.clipped.fq
207
   ln -s ../../fastq-files/SRR424707_2.fastq L1.R2.clipped.fq
   ln -s ../../fastq-files/SRR424708_1.fastq L2.R1.clipped.fq
```

```
210 ln -s ../../fastq-files/SRR424708_2.fastq L2.R2.clipped.fq
211 ln -s ../../fastq-files/SRR424709_1.fastq L3.R1.clipped.fq
212 ln -s ../../fastq-files/SRR424709_2.fastq L3.R2.clipped.fq
213
214 cd ../../ #returning to the output directory
```

215 3.3 Downloading genomic data from D. melanogaster

Now that we have the fastq files from the RAMPAGE libraries downloaded and named appropriately, we now must retrieve the genome assembly and rRNA sequences from *D. melanogaster*. The genome assembly is required for aligning the RAMPAGE reads, and the rRNA sequences are required to filter out matching reads in the sequenced RAMPAGE libraries. Because our sample is intended to contain only capped RNAs, any rRNA sequences we observe in these RAMPAGE libraries are contaminants that must be removed.

223

Please download the rRNA sequences from the demo/additional_files folder in the demo. These sequences were retrieved separately from Genbank at the NCBI database. Please navigate to the rRNA file "Dmel_rRNA.fasta" found in the Demo.

228 head -n 3

231 ACTAACAAGGATTTTCTTAGTAGCGGCGAGCGAAAAGAAACAGTTCAGCACTAAGTCACTTTGTCTATA

We will then download a version of the *D. melanogaster* genome assembly from ENSEMBL (www.ensembl.org) [29]. To retrieve the genome assembly, please do the following:

```
mkdir genome

cd genome

wget ftp://ftp.ensembl.org/pub/release-78/fasta/drosophila_melanogaster/dna/Drosophila_m

#uncompressing the file

gzip -d Drosophila_melanogaster.BDGP5.dna.toplevel.fa.gz

cd ..
```

3.4 Filtering and alignment of RAMPAGE reads using GoRAMPAGE

At this stage we are ready to commence with the rRNA filtering and alignment of the RAMPAGE libraries. We will use GoRAMPAGE, a tool we developed, to perform these tasks in a concerted workflow. GoRAMPAGE runs TagDust [24] to remove rRNA and low-complexity reads and STAR [26] to align RAMPAGE (or other paired-end) reads to a given genome assembly.

Setting up the GoRAMPAGE job. Please refer to the script "GoRAMPAGE_script_MMB.sh" and (using a text editor) provide the appropriate paths to the genome assembly, output directory (see above) and rRNA sequences (see Note 4). GoRAMPAGE jobs can optionally be run in parallel (see Note 5). The script can be executed as follows:

#vi GoRAMPAGE_script_MMB.sh #updating with a text editor
// CoRAMPAGE_script_MMB.sh

If everything is working correctly you should start to see the results of the job being written to the file "errScript". You can inspect the progress during the run using the *less* command.

258 less -S errScript

Should the run fail before completion, any associated error messages will be printed to the errScript file. Once the job is complete, you should see the message "GoRAMPAGE job is complete!" appear on the command-line terminal.

Inspecting the rRNA filtering results. To evaluate the results from Step 262 3 (rRNA filtering), please navigate to the top level of the "output" directory and open the file "LOGFILES". You'll see the recorded progress of the program Tagdust and a record of the results. We notice that (for the L3h library) 1046448 of reads (78.1%) were "extracted", meaning that slightly more than 20% of 266 reads were removed because of matches with ribosomal sequences. The removed 267 reads from all libraries are found in the "dusted_discard" directory, and the extracted reads are found in the current directory. Due to their sheer abundance 269 within cells, ribosomal RNA sequences are an inevitable contaminant within TSS 270 profiling libraries. For analysis purposes, it is important that these sequences be 271 removed, which is what has been completed here.

273 Since this step was conducted appropriately, we can proceed to the next step.

Evaluating the alignments. The folder "alignments/" in your GoRAMPAGE 274 output folder will now contain 6 bam files, each representing the distinct RAM-275 PAGE libraries selected for our analysis. Typing "ls -l" from the command line 276 will show that these files are symlinks to the original alignment files found in the "STARoutput/" directory. "STARoutput/", as its name suggests, con-278 tains the output from the STAR alignment, and this includes the alignment files 279 "*.sortedByCoord.out.bam", and four additional log files. The files with the suffix "*.STAR.Log.final.out" each contain a summary of the alignment, such as the number of input reads, the percentage of uniquely-mapped reads and the 282 percentage of unmapped reads. An inspection of these log files indicates that 283 the alignments have similar mapping rates (70-80%), a reasonable outcome for our purposes. 285

Now that our RAMPAGE libraries are filtered and aligned, we can commence with the second half of our analysis.

Promoter identification from aligned RAMPAGE libraries

We can now use the prepared alignment files to identify TSSs and promoters from 290 the selected RAMPAGE libraries. There are currently several tools available 291 for this purpose. CAGEr, developed by Haberle [30], was utilized to perform TSS identification as part of the FANTOM5 efforts. We will use TSRchitect in this demonstration, since it was specifically designed to analyze paired-end TSS 294 profiling datasets, and also because it is more flexible with respect to model 295 system (i.e. it does not require a corresponding BSGenome package). The latter feature will be helpful when analyzing the non-D. melanagaster TSS profiling 297 datasets that we expect to be generated in the near future.

Setting up the Analysis. TSRchitect, the package we'll use for this analysis, is an R package available in the Bioconductor suite of genomics tools [28]. 300 It makes use of existing packages and data structures within this environment, 301 where available, to identify promoters from sequence alignments. Since you have 302 already installed *TSRchitect* and its dependencies (see section 2.3), we are set 303 to proceed. 304

There are two general ways one can choose to run TSRchitect. The first is interactively i.e. typing the instructions directly into an R console. While this is a perfectly acceptable way to run analyses using package, for larger jobs 308 it will likely be more efficient (and likely more reproducible) to run a dedi-309 cated R script. We have provided a sample script "MMB_chapter_TSRchitect.R" 310 to make it easier for you to set up an R script. In the section to follow, we will go through the output of the analysis. For further details on how to use 312 TSRchitect, please see its documentation at its Bioconductor page found here: 313 https://www.bioconductor.org/packages/release/bioc/html/TSRchitect.html.

Running the Analysis. To run TSRchitect using the batch script, provide 315 full paths for the variables "BAMDIR" and "DmAnnot" in the script provided 316 (see Note 6). BAMDIR should be a path to the subdirectory "alignments/" in 317 RAMPAGE output directory you specified earlier, and DmAnnot should be a full path to the *D. melanogaster* gene annotation listed above. 319

Once this is complete, we can run the batch script from the Linux command-line as follows:

R CMD BATCH MMB_chapter_TSRchitect.R #assumes variables BAMDIR and DmAnnot have already been set 324 bg #puts this job in the background 325

Once the job is underway, you can monitor its progress by looking at the contents 326 of the .Rout file (in this case, "MMB_chapter_TSRchitect.Rout").

```
Reviewing the TSRchitect script. Before we evaluate the results (which
328
    will have been written to your working directory after running the batch script),
329
    there are some important aspects of the analysis to review. We discuss these for
    informational purposes only; it will not necessary to perform these commands
331
    separate from the batch script provided. First, we must initialize the tssObject
332
    (which stores the information about the experiment) appropriately (see Note 7).
    The inputs in this case are BAM files (inputType="bam"); TSRchitect also ac-
335
    cepts input in BED format.
336
    DmRAMPAGE <- loadTSSobj(experimentTitle = "RAMPAGE Tutorial", \</pre>
337
     inputDir=BAMDIR, inputType="bam", isPairedEnd=TRUE, \
     sampleNames=c("E1h","E2h", "E3h", "L1", "L2", "L3"), \
     replicateIDs=c(1,1,1,2,2,2))
    A critical step in our analysis is identifying TSRs from the aligned TSS data;
341
    to do this we use the function determine TSR. We have selected the job to run
    on 4 cores in this example (n.cores=4). Please enter the number of cores ap-
    propriate for your system. Because we want to identify TSRs from every one
    of the selected RAMPAGE libraries, we specify tssSet="all". The parameter
345
    tagCountThreshold was set to 25, meaning that only TSSs supported by 25 or
    more 5' RAMPAGE reads will be included within a TSR. Setting write Table to
    "TRUE" means that the identified TSRs from each set will be written to the
    working directory.
    DmRAMPAGE <- determineTSR(experimentName=DmRAMPAGE, n.cores=4, \</pre>
     tsrSetType="replicates", tssSet="all", tagCountThreshold=25, \
     clustDist=20, writeTable=TRUE)
       TSRchitect can incorporate the tag abundances from each of the samples
    and append them to the list of identified TSRs. This is useful for downstream
354
    analysis of differential expression.
355
    DmRAMPAGE <- addTagCountsToTSR(experimentName=DmRAMPAGE, \</pre>
    tsrSetType="replicates", tsrSet=1, tagCountThreshold=10, \
     writeTable=TRUE)
358
       We can use TSRchitect to import an annotation file (or, alternatively, use an
359
    existing one from AnnotationHub) and use it to associate our set of identified
    TSRs with coding genes. We can specify the maximum distances (both up-
    and downstream) between the TSR and the annotation using the arguments
    upstreamDist and downstreamDist.
363
    DmRAMPAGE <- importAnnotationExternal(experimentName=DmRAMPAGE, \</pre>
364
     fileType="gff3", annotFile=DmAnnot)
365
    DmRAMPAGE <- addAnnotationToTSR(experimentName=DmRAMPAGE, \
     tsrSetType="replicates", tsrSet=1, \
    upstreamDist=1000, downstreamDist=200, feature="gene", \
     featureColumnID="ID", writeTable=TRUE)
```

```
Now we have generated a set of identified TSSs, TSRs from all 6 RAMPAGE
371
    libraries, and have associated the identified TSRs with annotated genes. Next, we
372
    will merge the libraries into two samples according to condition: early embryonic
    (E1h, E2h, E3h) and late larval (L1, L2, L3) using the information we provided
374
    when we initialized the tssObject at the start of this section. After merging, we
    identify promoters i) within the merged samples and ii) within the entire dataset
    combined, and associate with the D. melanogaster gene annotation as described
    previously (not shown).
378
    #merging the sample data into two groups
379
    DmRAMPAGE <- mergeSampleData(DmRAMPAGE)</pre>
    # ... identifying TSRs from the merged samples:
382
    DmRAMPAGE <- determineTSR(experimentName=DmRAMPAGE, \</pre>
383
    n.cores=4, tsrSetType="merged", \
     tssSet="all", tagCountThreshold=40, \
     clustDist=20, writeTable=TRUE)
    Evaluating the results Our analysis using TSRchitect is now complete. Your
387
    working directory should now contain the following:
388
     - TSSs from each sample e.g. TSSset-1.txt: (6)
     - TSRs from each sample (in both .txt and .tab formats): (12)
     - TSRs from each merged group (in both .txt and .tab formats): e.g. TSRsetMerged-
        1.txt: (4)
392
     - TSRs from the combined set of TSSs: TSRsetCombined.tab: (1)
393
    Let's briefly review the files (see Note 8). We can quickly obtain the counts on
    the command line, as follows:
    wc -1 *.tab
    8377 TSRset-1.tab
397
    6159 TSRset-2.tab
    4814 TSRset-3.tab
    17924 TSRset-4.tab
    11851 TSRset-5.tab
401
    3242 TSRset-6.tab
402
    13986 TSRsetCombined.tab
    7344 TSRsetMerged-1.tab
    12126 TSRsetMerged-2.tab
405
    85823 total
    We will see that we have identified between roughly 3,200 and 18,000 TSRs
    within the individual RAMPAGE samples, which is attributable to the dif-
    ferences in library sizes. We detect 7,344 TSRs within the early embryonic
409
    samples ("TSRsetMerged-1.tab") and 12,126 TSRs in the late larval samples
410
    ("TSRsetMerged-2.tab"). Within the combined samples ("TSRsetCombined.tab")
```

we find 13,986 TSRs, which is similar to the number reported by Hoskins *et. al.* [1].

414

In addition to identifying the position of a given TSRs, *TSRchitect* records other useful information about its properties. The *width* of a TSR refers the span of the genomic region it occupies (in bp), and the *Shape Index* (SI) is measure of the relative peakedness of the TSR. We can see an example of this in the file "TSRsetMerged-1.txt".

420	seq	start	end	strand	nTSSs	tsrWidt	h	shapeIn	.dex	feature	:ID
421	2L.67043	3.67044.	+	2L	67043	67044	+	270	2	1	NA
422	2L.74089	9.74115.	+	2L	74089	74115	+	341	27	0.13	NA
423	2L.94739	9.94752.	+	2L	94739	94752	+	1650	14	0.55	FBgn0031
424	2L.10238	86.10238	6.+	2L	102386	102386	+	284	1	2	FBgn0031

425 3.6 Summary

The workflow provided here is intended to serve as a useful entry point for the 126 analysis of TSS profiling data in insects. On the computational side, we have 427 provided an open source set of tools so that the uninitiated genome scientist 428 can begin to analyze RAMPAGE (or other forms of TSS profiling data) quickly. While the analysis centered on *D. melanogaster* via the use of public datasets, it is anticipated that this will assist groups who may be interested in performing TSS profiling in their preferred insect model system. The application of TSS 432 profiling technology across a more representative sample of insect diversity will 433 improve our understanding of the positions and general structure cis-regulatory 434 regions in this phylum.

436 3.7 Figures

137 4 Notes

- 1. Please consult the GoRAMPAGE documentation found here:
 https://github.com/BrendelGroup/GoRAMPAGE.
 Installation instructions for the prerequisites of GoRAMPAGE (which includes some of the items listed) are found at the following link:
 https://github.com/BrendelGroup/GoRAMPAGE/tree/master/src.
- 2. You can clone the entire GoRAMPAGE repository (which includes the contents of the Demo) to your workspace on the command line using git, as follows:

```
git clone https://github.com/brendelgroup/GoRAMPAGE/
```

The "scripts/" folder in the Demo contains code for you to run the two major workflows described in this chapter. The "additional_files/" folder contains the following files which are necessary for the analysis: i) a fasta file containing ribosomal RNA sequences for D. melanogaster (Dmel_rRNA.fasta)

and ii) a gene annotation for D. melanogaster (Drosophila_melanogaster.BDGP5.78.gff).



Fig. 1. A brief summary of the RAMPAGE protocol. Starting with high-quality total RNA, first-strand cDNA synthesis is initiated using a cap-bound oligonucleotide and a custom RAMPAGE RT primer, creating a double-stranded DNA-RNA hybrid molecule. Next, the 5'-m7G cap is oxidized, bound with biotin and pulled down with streptavadin-coated beads. The single-stranded cDNA molecules is released and the final RAMPAGE library construction is completed with PCR using custom oligonucleotides, followed by size-selection. This illustration was adapted from [18].



Fig. 2. An overview of promoter identification using RAMPAGE. a) RAMPAGE reads are aligned to the genome. The 5'-most genomic coordinate from each properly-paired R1 read is estimated as a TSS. The ambundance of mapped 5'-ends at a given TSS is a measure of its abundance. TSSs above a minimum threshold will be clustered into TSRs. b) RAMPAGE-derived Paired-end sequence information provides a connection between a 5'-mRNA end and a gene coding region. Only properly-paired R1 reads (i.e. with an aligned R2 read) are identified as TSSs and then included in the downstream clustering procedure described in part a.

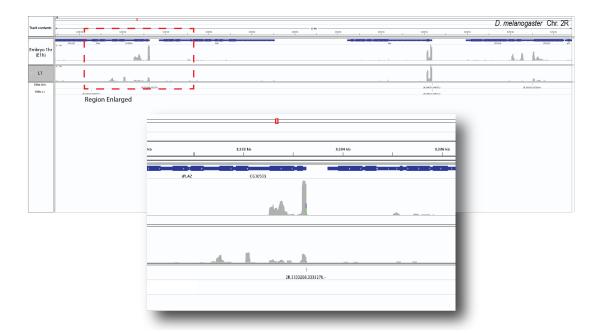


Fig. 3. An overview of the TSS profiling information provided by RAMPAGE. A representative visualization of RAMPAGE peaks (i.e. clusters of properly-aligned RAMPAGE reads) within an arbitrarily-selected genomic region of D. melanogaster chromosome 2R is shown, along with the corresponding gene annotation within this region. RAMPAGE data from two RAMPAGE libraries from Batut et al [2] are shown, which were generated from RNA isolated from developmental stages E1h and L1 see Methods. For each library, the abundance of RAMPAGE reads that align to a given site within the genome is represented by density plots (shown in gray). Gene models are shown in blue, where the thickened line represents exons and thin lines represent introns. The locations of TSRs identified by TSRchitect are shown in the two tracks from the bottom of the image. In some cases the expression of 5'-ends is roughly equivalent between the two samples, whereas in others the observed signal is substantially higher (see Inset). The original images are screenshots generated in the Integrated Genomic Viewer (IGV; http://software.broadinstitute.org/software/igv/) [31]. Where necessary, additional annotation was added using Adobe Illustrator.

- 3. Since these fastq files are paired-end, we use the argument *-split-files* to generate separate files for each read pair.
- 455 4. If you are running this on a cluster with a job scheduler you'll need to add
 456 the necessary headers to the top of the script and submit the job in the
 457 appropriate manner.
- 5. For parallel execution, GoRAMPAGE uses the Linux package *GNU parallel* [32]. Please see the GoRAMPAGE documentation for more information.
- 6. To do this, please edit the batch script TSRchitect_script_MMB.R with a text editor of your choice.
- 7. Because the samples provided derive from related developmental stages, we will merge them for annotation purposes using the argument *replicateIDs*, (though it must be emphasized that they are not replicates).
- 8. All of *TSRchitect's* output files are labeled according to the order that they are loaded onto the *tssObject*. For example, *TSSset-1.txt* corresponds to the first RAMPAGE dataset (in our case E1h), and *TSSset-2.txt* corresponds to the second RAMPAGE dataset (for this example E2h), and so on. You can check which datasets are loaded on the *tssObject* by simply entering it on an R console. Please see the *TSRchitect* documentation for more information.

Acknowledgments

The authors would like to thank Philippe Batut for generous technical assistance with the RAMPAGE protocol, and to Nathan Keith for his help establishing the protocol in our laboratory.

Disclosure Declaration

The authors declare that they have no competing interests.

477 5 References

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633 6 Checklist of Items to be Sent to Volume Editors

534	Here is a checklist of everything the volume editor requires from you:
535	☐ The final LATEX source files
536	☐ A final PDF file

	2	Raborn and Brendel
63 63		A copyright form, signed by one author on behalf of all of the authors of the paper.
63	9 [A readme giving the name and email address of the corresponding author.