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

Raborn and Brendel

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).

17

2

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 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.

1.4 Paired-end TSS Profiling with RAMPAGE

55

83

85

89

90

92

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 (see Figure 1). As with CAGE 73 and other TSS profiling methods, RAMPAGE reads are aligned, to obtain TSSs 74 and clustered to identify Transcription Start Regions (TSRs), which are en-75 richments of TSSs consistent with promoters (Figure 2A). A key advantage of generating paired-end sequence is transcript connectivity, which provides a di-77 rect link between a given 5'-end and its associated mRNA molecule [2] (Figure 78 2B). Because short or spurious RNAs are found within the transcriptome, tran-79 script connectivity allows the TSSs (and thus promoters) of full-length mRNAs 80 to be unambiguously identified, which benefits genome annotation and improves 81 interpretation of transcript species. 82

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

4 Raborn and Brendel

- from the aligned sequences and associate them with coding regions. In closing,
- 97 we will consider further applications of this data and discuss the utility of re-
- 98 producible workflows in bioinformatic analysis.

99 2 Materials

The example analyses described herein require a workstation capable of doing 100 modern bioinformatics; minimally a reasonably-appointed laptop. An interme-101 diate understanding of the Linux/Unix command line will be extremely useful, 102 although we make efforts to explain the procedures with clarity. In addition, it 103 will likely be necessary for the participant to have superuser privileges on the machine. If you do not have a machine (or have access to one) that meets these 105 requirements, it is recommended that you consider cloud-based cyberinfrastruc-106 ture, including Amazon Web Services (AWS; https://aws.amazon.com/), Cy-107 Verse (http://www.cyverse.org/) [19], or JetStream (https://jetstream-cloud.org/) [20]. The former is a well-known pay-per-use solution, while the latter two are 109 NSF-funded resources that makes compute allocations freely available to the 110 public. 111

112 2.1 Hardware

- 1. x86-64 compatible processors
- 114 2. 16GB RAM
- 3. 80GB+ hard disk space

116 2.2 Operating System

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

119 2.3 Software

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

123 Sequence retrieval

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

GoRAMPAGE

127

- 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)

132 TSRchitect

```
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**)

137 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 here:
- https://github.com/brendelgroup/GoRAMPAGE/demo/MMB (see Note 2).

142 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] (see Note 3). To install Bioconductor, please type the following from an R console:
```

```
source("https://bioconductor.org/biocLite.R")
biocLite()
```

We will use the R package *TSRchitect* to identify promoters from aligned RAM-PAGE libraries. Prior to running the analysis, it will be necessary to install a 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",
"ENCODExplorer", "GenomicAlignments", "GenomeInfoDb",
"GenomicRanges", "IRanges", "methods",
"Rsamtools", "rtracklayer", "S4Vectors",
"SummarizedExperiment"))

biocLite("TSRchitect")
```

Finally, please confirm that TSRchitect has been installed correctly by loading it from your R console as follows:

```
library(TSRchitect) #loading TSRchitect
```

164 3 Methods

55 3.1 Retrieving the RAMPAGE sequence data from NCBI

```
To begin our analysis, we must download the RAMPAGE data to our worksta-
    tion. We will utilize tools provided by the SRA Toolkit, which should already
167
    be installed on your machine (see Materials). The command fastq-dump al-
168
    lows one to directly retrieve data from the GEO database using the appropriate
169
    identifier(s). While there are 36 RAMPAGE libraries in the Batut et al. pa-
    per, we will select a subset of these to analyze here. We will compare samples
171
    from selected embryonic (E01h-E03h) and larval (L1-L3) tissues, representing
172
    the beginning and end of embryonic development. For more information about
    the experiment and the available RAMPAGE libraries, please see the following
    link: https://www.ncbi.nlm.nih.gov/Traces/study/?acc=SRP011193.
175
176
    First, let's proceed with downloading the libraries from early embryonic tissues
    (see See Note 4). We will make a new folder (entitled "fastq_files/") to
    house these files.
179
    mkdir fastq_files
180
    cd fastq_files
181
    fastq-dump --split-files SRR424683
    fastq-dump --split-files SRR424684
184
    fastq-dump --split-files SRR424685
    We continue by downloading the data from late larval tissues.
    fastq-dump --split-files SRR424707
187
    fastq-dump --split-files SRR424708
188
    fastq-dump --split-files SRR424709
    Once the download of the aforementioned files are complete, you should see a
190
    total of 12 (6 x 2) separate fastq files in your current working directory:
    ls -l *.fastq | wc -l
         Creating symlinks to the files
193
    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-
    links) to the files, as follows:
    cd ..
    mkdir -p output/reads/clipped
    cd output/reads/clipped
199
    #embryonic libraries
```

```
ln -s ../../fastq-files/SRR424683_1.fastq E01h.R1.clipped.fq
202
   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
   ln -s ../../fastq-files/SRR424685_2.fastq E03h.R2.clipped.fq
   #larval libraries
209
   ln -s ../../fastq-files/SRR424707_1.fastq L1.R1.clipped.fq
210
   ln -s ../../fastq-files/SRR424707_2.fastq L1.R2.clipped.fq
   ln -s ../../fastq-files/SRR424708_1.fastq L2.R1.clipped.fq
   ln -s ../../fastq-files/SRR424708_2.fastq L2.R2.clipped.fq
213
   ln -s ../../fastq-files/SRR424709_1.fastq L3.R1.clipped.fq
214
   ln -s ../../fastq-files/SRR424709_2.fastq L3.R2.clipped.fq
216
   cd ../../.. #returning to the output directory
217
```

218 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.

226 227

228

240

24

Please make note of the rRNA sequences, found in the file "Dmel_rRNA.fasta", from the folder additional_files folder in the demo (see Note 5).

229

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

cd genome

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

drosophila_melanogaster/dna/Drosophila_melanogaster.BDGP5.dna.toplevel.fa.gz

"uncompressing the file

zas 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

./GoRAMPAGE_script_MMB.sh

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 247 "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 6). GoRAMPAGE jobs can optionally be run in parallel (see Note 250 **7**). The script can be executed as follows: 251 #vi GoRAMPAGE_script_MMB.sh #updating with a text editor 252

If everything is working correctly you should start to see the results of the job 254 being written to the file "errScript". You can inspect the progress during the run using the *less* command.

less -S errScript 257

256

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 259 "Gorampage job is complete!" appear on the command-line terminal.

Inspecting the rRNA filtering results. To evaluate the results from Step 261 3 (rRNA filtering), please navigate to the top level of the "output" directory 262 and open the file "LOGFILES". You'll see the recorded progress of the program 263 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 265 reads were removed because of matches with ribosomal sequences. The removed 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 within cells, ribosomal RNA sequences are an inevitable contaminant within TSS 269 profiling libraries. For analysis purposes, it is important that these sequences be 270 removed, which is what has been completed here. Since this step was conducted appropriately, we can proceed to the next step. 272

Evaluating the alignments. The folder "alignments/" in your GoRAMPAGE 273 output folder will now contain 6 .bam files, each representing the distinct RAM-PAGE libraries selected for our analysis. Typing "ls-l" from the command line will show that these files are symlinks to the original alignment files found 276 in the "STARoutput/" directory. "STARoutput/", as its name suggests, con-277 tains the output from the STAR alignment, and this includes the alignment files "*.sortedByCoord.out.bam", and four additional log files. The files with the suf-279 fix "*.STAR.Log.final.out" each contain a summary of the alignment, such as 280 the number of input reads, the percentage of uniquely-mapped reads and the percentage of unmapped reads. An inspection of these log files indicates that

the alignments have similar mapping rates (~70-80%), a reasonable outcome for 283 our purposes. 284

Now that our RAMPAGE libraries are filtered and aligned, we can commence 286 with the second half of our analysis.

Promoter identification from aligned RAMPAGE libraries 3.5288

We can now use the prepared alignment files to identify TSSs and promoters from the selected RAMPAGE libraries. There are currently several tools avail-290 able for this purpose. CAGEr, developed by Haberle [30], was utilized to perform 291 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 profiling datasets, and also because it is more flexible with respect to model sys-294 tem (i.e. it does not require a corresponding BSGenome [31] package). The latter 295 feature will be helpful when analyzing the non-D. melanagaster TSS profiling datasets that we expect to be generated in the near future. 297

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]. It makes use of existing packages and data structures within this environment, where available, to identify promoters from sequence alignments. Since you have already installed *TSRchitect* and its dependencies (see section 2.3), we are set to proceed.

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 it will likely be more efficient (and likely more reproducible) to run a dedicated R script. We have provided sample scripts to make it easier for you to set up an R script. The two scripts are identical with a single exception: one is set up to run in parallel ("TSRchitect_parallel_MMB.R"), while the other is written to run in serial ("TSRchitect_serial_MMB.R"). Please select the script that best suits your computing resources. In the section to follow, we will go through the output of the analysis. For further details on how to use TSRchitect, please see its documentation at its Bioconductor page found here: https://www.bioconductor.org/packages/release/bioc/html/TSRchitect.html.

Running the Analysis. To run TSRchitect using the batch script, provide 317 full paths for the variables "BAMDIR" and "DmAnnot" in the script provided (see Note 8). BAMDIR should be a path to the subdirectory "alignments/" in 319 RAMPAGE output directory you specified earlier, and DmAnnot should be a 320 full path to the *D. melanogaster* gene annotation listed above. 321

298

299

301

302

305

306

308

309

310

312

313

314

316

```
as follows:
    R CMD BATCH TSRchitect_parallel_MMB.R #or use 'serial script
    #assumes variables BAMDIR and DmAnnot have already been set
326
    bg #puts this job in the background
327
    Once the job is underway, you can monitor its progress by looking at the contents
328
    of the .Rout file (in this case, "TSRchitect_parallel_MMB.Rout").
329
    Reviewing the TSRchitect script. Before we evaluate the results (which
330
    will have been written to your working directory after running the batch script),
331
    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
    separate from the batch script provided. First, we must initialize the tssObject
    (which stores the information about the experiment) appropriately (see Note 9).
335
336
    The inputs in this case are BAM files (inputType="bam"); TSRchitect also ac-
337
    cepts input in BED format.
    DmRAMPAGE <- loadTSSobj(experimentTitle = "RAMPAGE Tutorial", \</pre>
339
     inputDir=BAMDIR, inputType="bam", isPairedEnd=TRUE, \
     sampleNames=c("E1h","E2h", "E3h", "L1", "L2", "L3"), \
     replicateIDs=c(1,1,1,2,2,2))
342
    A critical step in our analysis is identifying TSRs from the aligned TSS data;
343
    to do this we use the function determine TSR. We have selected the job to run
344
    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
347
    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.
351
    DmRAMPAGE <- determineTSR(experimentName=DmRAMPAGE, n.cores=4, \
352
     tsrSetType="replicates", tssSet="all", tagCountThreshold=25, \
353
     clustDist=20, writeTable=TRUE)
    TSRchitect can incorporate the tag abundances from each of the samples and
355
    append them to the list of identified TSRs. This is useful for downstream analysis
    of differential expression.
357
    DmRAMPAGE <- addTagCountsToTSR(experimentName=DmRAMPAGE, \</pre>
358
    tsrSetType="replicates", tsrSet=1, tagCountThreshold=10, \
     writeTable=TRUE)
```

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

```
We can use TSRchitect to import an annotation file (or, alternatively, use an
    existing one from AnnotationHub) and use it to associate our set of identified
362
    TSRs with coding genes. We can specify the maximum distances (both up-
    and downstream) between the TSR and the annotation using the arguments
364
    upstreamDist and downstreamDist.
    DmRAMPAGE <- importAnnotationExternal(experimentName=DmRAMPAGE, \</pre>
     fileType="gff3", annotFile=DmAnnot)
367
    DmRAMPAGE <- addAnnotationToTSR(experimentName=DmRAMPAGE, \</pre>
     tsrSetType="replicates", tsrSet=1, \
    upstreamDist=1000, downstreamDist=200, feature="gene", \
371
     featureColumnID="ID", writeTable=TRUE)
372
    Now we have generated a set of identified TSSs, TSRs from all 6 RAMPAGE
373
    libraries, and have associated the identified TSRs with annotated genes. Next, we
    will merge the libraries into two samples according to condition: early embryonic
375
    (E1h, E2h, E3h) and late larval (L1, L2, L3) using the information we provided
376
    when we initialized the tssObject at the start of this section. After merging, we
377
    identify promoters i) within the merged samples and ii) within the entire dataset
    combined, and associate with the D. melanogaster gene annotation as described
379
    previously (not shown).
    #merging the sample data into two groups
    DmRAMPAGE <- mergeSampleData(DmRAMPAGE)</pre>
382
    # ... identifying TSRs from the merged samples:
    DmRAMPAGE <- determineTSR(experimentName=DmRAMPAGE, \</pre>
    n.cores=4, tsrSetType="merged", \
386
     tssSet="all", tagCountThreshold=40, \
387
     clustDist=20, writeTable=TRUE)
    Evaluating the results Our analysis using TSRchitect is now complete. A
    snapshot of a representative sample of small set of aligned RAMPAGE libraries
    is shown in Figure 3. Your working directory should now contain the following:
391
     - TSSs from each sample e.g. TSSset-1.txt: (6)
392
     - 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)
395
     - TSRs from the combined set of TSSs: TSRsetCombined.tab: (1)
    Let's briefly review the files (see Note 10). We can quickly obtain the counts
    on the command line, as follows:
398
    wc -1 *.tab
    8377 TSRset-1.tab
```

```
6159 TSRset-2.tab
401
    4814 TSRset-3.tab
402
    17924 TSRset-4.tab
    11851 TSRset-5.tab
404
   3242 TSRset-6.tab
    13986 TSRsetCombined.tab
    7344 TSRsetMerged-1.tab
    12126 TSRsetMerged-2.tab
408
    85823 total
```

We will see that we have identified between roughly 3,200 and 18,000 TSRs 410 within the individual RAMPAGE samples, which is attributable to the differences in library sizes. We detect 7,344 TSRs within the early embryonic 412 samples ("TSRsetMerged-1.tab") and 12.126 TSRs in the late larval samples 413 ("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].416

417

409

In addition to identifying the position of a given TSRs, TSRchitect records other 418 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 420 the relative peakedness of the TSR. We can see an example of this in the file 421 "TSRsetMerged-1.txt". 422

423	seq	start	end	strand	nTSSs	tsrWidtl	n	shapeIn	dex	feature	ID
424	2L.67043	3.67044.	+	2L	67043	67044	+	270	2	1	NA
425	2L.74089	9.74115.	+	2L	74089	74115	+	341	27	0.13	NA
426	2L.94739	9.94752.	+	2L	94739	94752	+	1650	14	0.55	FBgn0031
427	2L.10238	36.10238	6.+	2L	102386	102386	+	284	1	2	FBgn0031

3.6 Summary 428

The workflow provided here is intended to serve as a useful entry point for the 429 analysis of TSS profiling data in insects. On the computational side, we have provided an open source set of tools so that the uninitiated genome scientist 431 can begin to analyze RAMPAGE (or other forms of TSS profiling data) quickly. 432 While the analysis centered on D. melanogaster via the use of public datasets, 433 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 435 profiling technology across a more representative sample of insect diversity will 436 improve our understanding of the positions and general structure cis-regulatory 437 regions in this phylum.



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].

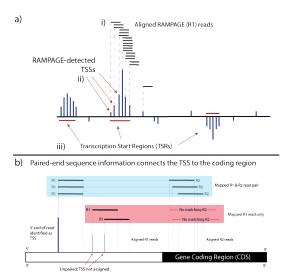


Fig. 2. An overview of promoter identification using RAMPAGE. a) RAMPAGE reads are aligned to the genome. The 5'-most genomic coordinate from each properly-paired R1 read is estimated as a TSS. The abundance of mapped 5'-ends at a given TSS is a measure of its abundance. TSSs above a minimum threshold 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.

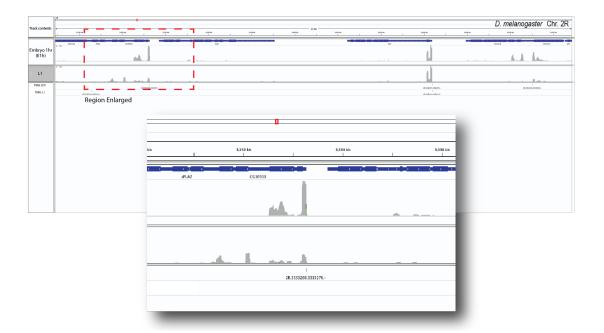


Fig. 3. An overview of the TSS profiling information provided by RAMPAGE. A representative visualization of RAMPAGE peaks (i.e. clusters of properly-aligned RAM-PAGE reads) within an arbitrarily-selected genomic region of D. melanogaster chromosome 2R is shown, along with the corresponding gene annotation 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. A single region, highlighted with the red dashed line is enlarged (the *Inset*) to show further detail of a selected gene and RAMAPGE signals. In some cases, the expression of 5'-ends between the two samples is roughly equivalent, 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/) [32]. Where necessary, additional annotation was added using Adobe Illustrator.

470

439 4 Figures

440 5 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. On Linux, the installation of a few packages are necessary in order to install Bioconductor packages using biocLite().
- To install them using Ubuntu:

```
apt-get install libssl-dev
apt-get install libcurl4-openssl-dev
apt-get install libxml2-dev
```

- If you do not Ubuntu, use the commands necessary to install the above packages on your Linux distribution.
- 3. 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/
cd demo/MMB
```

- 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).
- 465 4. Since these fastq files are paired-end, we use the argument -split-files to generate separate files for each read pair.
- 5. If you are running this on a cluster with a job scheduler you'll need to add the necessary headers to the top of the script and submit the job in the appropriate manner.
 - 6. The rRNA sequences were retrieved separately from Genbank at NCBI [33].
- 7. For parallel execution, GoRAMPAGE uses the Linux package *GNU parallel* [34]. Please see the GoRAMPAGE documentation for more information.
- 8. To do this, please edit the batch script TSRchitect_serial_MMB.R with a text editor of your choice.
- 9. 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).

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.

488 Disclosure Declaration

The authors declare that they have no competing interests.

490 6 References

491 References

484

- R. A. Hoskins, R. A. Hoskins, J. M. Landolin, J. M. Landolin, J. B. Brown, J. B. Brown, J. E. Sandler, J. E. Sandler, H. Takahashi, H. Takahashi, T. Lassmann, T. Lassmann, C. Yu, C. Yu, B. W. Booth, B. W. Booth, D. Zhang, D. Zhang, K. H. Wan, K. H. Wan, L. Yang, L. Yang, N. Boley, N. Boley, J. Andrews, J. Andrews, T. C. Kaufman, T. C. Kaufman, B. R. Graveley, B. R. Graveley, P. J. Bickel, P. J. Bickel, P. Carninci, J. W. Carlson, J. W. Carlson, S. E. Celniker, and S. E. Celniker, "Genome-wide analysis of promoter architecture in Drosophila melanogaster." Genome Research, vol. 21, no. 2, pp. 182–192, Feb. 2011.
- P. J. Batut, A. Dobin, C. Plessy, P. Carninci, and T. R. Gingeras, "High-fidelity promoter profiling reveals widespread alternative promoter usage and transposon-driven developmental gene expression." Genome Research, Aug. 2012.
- 3. V. P. Brendel and R. T. Raborn, GoRAMPAGE- A Workflow For Promoter Detection by 5'-Read Mapping, https://github.com/brendelGroup/GoRAMPAGE, 2016.
- 4. R. T. Raborn and V. Brendel, TSRchitect: Promoter identification from large-scale
 TSS profiling data, 2017, r Bioconductor package version 1.0.0. [Online]. Available:
 http://bioconductor.org/packages/release/bioc/html/TSRchitect.html
- 508 5. J. T. Kadonaga, "Perspectives on the RNA polymerase II core promoter." Wiley
 509 Interdisciplinary Reviews: Developmental Biology, vol. 1, no. 1, pp. 40–51, Jan.
 510 2012.
- 6. R. Kodzius, M. Kojima, H. Nishiyori, M. Nakamura, S. Fukuda, M. Tagami,
 D. Sasaki, K. Imamura, C. Kai, M. Harbers, Y. Hayashizaki, and P. Carninci,
 "CAGE: cap analysis of gene expression." Nature Methods, vol. 3, no. 3, pp. 211–
 222, Mar. 2006.
- 7. P. Carninci, T. Kasukawa, S. Katayama, J. Gough, M. C. Frith, N. Maeda,
 R. Oyama, T. Ravasi, B. Lenhard, C. Wells, R. Kodzius, K. Shimokawa, V. B.
 Bajic, S. E. Brenner, S. Batalov, A. R. R. Forrest, M. Zavolan, M. J. Davis, L. G.

Wilming, V. Aidinis, J. E. Allen, A. Ambesi-Impiombato, R. Apweiler, R. N. Atu-518 raliya, T. L. Bailey, M. Bansal, L. Baxter, K. W. Beisel, T. Bersano, H. Bono, A. M. 519 Chalk, K. P. Chiu, V. Choudhary, A. Christoffels, D. R. Clutterbuck, M. L. Crowe, 520 E. Dalla, B. P. Dalrymple, B. de Bono, G. Della Gatta, D. di Bernardo, T. Down, 521 P. Engstrom, M. Fagiolini, G. Faulkner, C. F. Fletcher, T. Fukushima, M. Furuno, 522 S. Futaki, M. Gariboldi, P. Georgii-Hemming, T. R. Gingeras, T. Gojobori, R. E. 523 Green, S. Gustincich, M. Harbers, Y. Hayashi, T. K. Hensch, N. Hirokawa, D. Hill, L. Huminiecki, M. Iacono, K. Ikeo, A. Iwama, T. Ishikawa, M. Jakt, A. Kanapin, M. Katoh, Y. Kawasawa, J. Kelso, H. Kitamura, H. Kitano, G. Kollias, S. P. T. Krishnan, A. Kruger, S. K. Kummerfeld, I. V. Kurochkin, L. F. Lareau, D. Lazarevic, L. Lipovich, J. Liu, S. Liuni, S. McWilliam, M. Madan Babu, M. Madera, L. Marchionni, H. Matsuda, S. Matsuzawa, H. Miki, F. Mignone, S. Miyake, K. Mor-529 ris, S. Mottagui-Tabar, N. Mulder, N. Nakano, H. Nakauchi, P. Ng, R. Nilsson, 530 S. Nishiguchi, S. Nishikawa, F. Nori, O. Ohara, Y. Okazaki, V. Orlando, K. C. 531 Pang, W. J. Pavan, G. Pavesi, G. Pesole, N. Petrovsky, S. Piazza, J. Reed, J. F. 532 Reid, B. Z. Ring, M. Ringwald, B. Rost, Y. Ruan, S. L. Salzberg, A. Sandelin, 533 C. Schneider, C. Schönbach, K. Sekiguchi, C. A. M. Semple, S. Seno, L. Sessa, 534 Y. Sheng, Y. Shibata, H. Shimada, K. Shimada, D. Silva, B. Sinclair, S. Sperling, 535 E. Stupka, K. Sugiura, R. Sultana, Y. Takenaka, K. Taki, K. Tammoja, S. L. Tan, 536 S. Tang, M. S. Taylor, J. Tegner, S. A. Teichmann, H. R. Ueda, E. van Nimwegen, 537 R. Verardo, C. L. Wei, K. Yagi, H. Yamanishi, E. Zabarovsky, S. Zhu, A. Zim-538 mer, W. Hide, C. Bult, S. M. Grimmond, R. D. Teasdale, E. T. Liu, V. Brusic, 539 J. Quackenbush, C. Wahlestedt, J. S. Mattick, D. A. Hume, C. Kai, D. Sasaki, 540 Y. Tomaru, S. Fukuda, M. Kanamori-Katayama, M. Suzuki, J. Aoki, T. Arakawa, 541 J. Iida, K. Imamura, M. Itoh, T. Kato, H. Kawaji, N. Kawagashira, T. Kawashima, 542 M. Kojima, S. Kondo, H. Konno, K. Nakano, N. Ninomiya, T. Nishio, M. Okada, 543 C. Plessy, K. Shibata, T. Shiraki, S. Suzuki, M. Tagami, K. Waki, A. Watahiki, 544 Y. Okamura-Oho, H. Suzuki, J. Kawai, Y. Hayashizaki, F. Consortium, R. G. E. R. 545 Group, and G. S. G. G. N. P. C. Group, "The transcriptional landscape of the mammalian genome," Science (New York, NY), vol. 309, no. 5740, pp. 1559-1563, Sep. 2005.

- 8. E. A. Rach, H.-Y. Yuan, W. H. Majoros, P. Tomancak, and U. Ohler, "Motif composition, conservation and condition-specificity of single and alternative transcription start sites in the Drosophila genome." *Genome Biology*, vol. 10, no. 7, p. R73, 2009.
- 9. B. Lenhard, A. Sandelin, and P. Carninci, "Metazoan promoters: emerging characteristics and insights into transcriptional regulation." Nature Reviews Genetics, vol. 13, no. 4, pp. 233–245, Apr. 2012.
- T. Ni, D. L. Corcoran, E. A. Rach, S. Song, E. P. Spana, Y. Gao, U. Ohler,
 and J. Zhu, "A paired-end sequencing strategy to map the complex landscape of transcription initiation." *Nature Methods*, vol. 7, no. 7, pp. 521–527, Jul. 2010.
- U. Ohler, G.-c. Liao, H. Niemann, and G. M. Rubin, "Computational analysis of core promoters in the Drosophila genome." Genome Biology, vol. 3, no. 12, pp. research0087.1–0087.12, 2002.
- 12. R. T. Raborn, K. Spitze, V. P. Brendel, and M. Lynch, "Promoter Architecture and Sex-Specific Gene Expression in Daphnia pulex." Genetics, vol. 204, no. 2, pp. 593–612, Aug. 2016.
- 13. C. Nepal, Y. Hadzhiev, C. Previti, V. Haberle, N. Li, H. Takahashi, A. M. M.
 Suzuki, Y. Sheng, R. F. Abdelhamid, S. Anand, J. Gehrig, A. Akalin, C. E. M.
 Kockx, A. A. J. van der Sloot, W. F. J. van IJcken, O. Armant, S. Rastegar,

- C. Watson, U. Strahle, E. Stupka, P. Carninci, B. Lenhard, and F. Muller, "Dynamic regulation of the transcription initiation landscape at single nucleotide resolution during vertebrate embryogenesis," *Genome Research*, vol. 23, no. 11, pp. 1938–1950, Nov. 2013.
- 14. P. Carninci, A. Sandelin, B. Lenhard, S. Katayama, K. Shimokawa, J. Ponjavic,
 C. A. M. Semple, M. S. Taylor, P. G. Engström, M. C. Frith, A. R. R. Forrest, W. B. Alkema, S. L. Tan, C. Plessy, R. Kodzius, T. Ravasi, T. Kasukawa,
 S. Fukuda, M. Kanamori-Katayama, Y. Kitazume, H. Kawaji, C. Kai, M. Nakamura, H. Konno, K. Nakano, S. Mottagui-Tabar, P. Arner, A. Chesi, S. Gustincich,
 F. Persichetti, H. Suzuki, S. M. Grimmond, C. A. Wells, V. Orlando, C. Wahlestedt, E. T. Liu, M. Harbers, J. Kawai, V. B. Bajic, D. A. Hume, and Y. Hayashizaki,
 "Genome-wide analysis of mammalian promoter architecture and evolution," Nature Genetics, vol. 38, no. 6, pp. 626–635, Apr. 2006.
- 15. S. Mwangi, G. Attardo, Y. Suzuki, S. Aksoy, and A. Christoffels, "TSS seq based core promoter architecture in blood feeding Tsetse fly (Glossina morsitans morsitans) vector of Trypanosomiasis," BMC Genomics, vol. 16, no. 1, p. 722, Sep. 2015.
- K. Tsuchihara, Y. Suzuki, H. Wakaguri, T. Irie, K. Tanimoto, S.-i. Hashimoto,
 K. Matsushima, J. Mizushima-Sugano, R. Yamashita, K. Nakai, D. Bentley, H. Esumi, and S. Sugano, "Massive transcriptional start site analysis of human genes in hypoxia cells," *Nucleic Acids Research*, vol. 37, no. 7, pp. 2249–2263, Apr. 2009.
- 17. N. Cvetesic and B. Lenhard, "Core promoters across the genome," *Nature Biotech-nology*, vol. 35, no. 2, pp. 123–124, Feb. 2017.
- 18. P. J. Batut and T. R. Gingeras, "RAMPAGE: Promoter Activity Profiling by Paired-End Sequencing of 5'-Complete cDNAs." in *Current Protocols in Molecular Biology*. Current protocols in molecular biology / edited by Frederick M Ausubel
 [et al], 2013, pp. 25B.11.1–25B.11.16.
- 19. N. Merchant, E. Lyons, S. Goff, M. Vaughn, D. Ware, D. Micklos, and P. Antin,
 "The iPlant Collaborative: Cyberinfrastructure for Enabling Data to Discovery for
 the Life Sciences." *PLoS Biology*, vol. 14, no. 1, p. e1002342, Jan. 2016.
- 20. C. A. Stewart, T. M. Cockerill, I. Foster, D. Hancock, N. Merchant,
 E. Skidmore, D. Stanzione, J. Taylor, S. Tuecke, G. Turner, M. Vaughn,
 and N. I. Gaffney, "Jetstream: A self-provisioned, scalable science and
 engineering cloud environment," in Proceedings of the 2015 XSEDE Conference:
 Scientific Advancements Enabled by Enhanced Cyberinfrastructure, ser. XSEDE
 New York, NY, USA: ACM, 2015, pp. 29:1–29:8. [Online]. Available:
 http://doi.acm.org/10.1145/2792745.2792774
- 21. R. Leinonen, H. Sugawara, M. Shumway, and International Nucleotide Sequence
 Database Collaboration, "The sequence read archive." Nucleic Acids Research,
 vol. 39, no. Database issue, pp. D19–21, Jan. 2011.
- E. Aronesty, "Comparison of Sequencing Utility Programs," The Open Bioinformatics Journal, vol. 7, no. 1, pp. 1–8, Jan. 2013.
- 610 23. H. Lab, "FASTX Toolkit." [Online]. Available: 611 http://hannonlab.cshl.edu/fastx_toolkit/
- 24. T. Lassmann, "TagDust2: a generic method to extract reads from sequencing data,"
 BMC Bioinformatics, vol. 16, no. 1, p. 1, Jan. 2015.
- 25. H. Li, B. Handsaker, A. Wysoker, T. Fennell, J. Ruan, N. Homer, G. Marth, G. R.
 Abecasis, R. Durbin, and 1000 Genome Project Data Processing Subgroup, "The
 Sequence Alignment/Map format and SAMtools," Bioinformatics (Oxford, England), vol. 25, no. 16, pp. 2078–2079, Aug. 2009.

- 26. A. Dobin and T. R. Gingeras, "Optimizing RNA-Seq Mapping with STAR," in
 Transcription Factor Regulatory Networks. New York, NY: Springer New York,
 Apr. 2016, pp. 245–262.
- 27. R Core Team, R: A Language and Environment for Statistical Computing, R
 Foundation for Statistical Computing, Vienna, Austria, 2017. [Online]. Available:
 https://www.R-project.org
- 28. M. Lawrence and M. Morgan, "Scalable Genomics with R and Bioconductor,"
 Statistical Science, vol. 29, no. 2, pp. 214–226, May 2014.
- A. Yates, W. Akanni, M. R. Amode, D. Barrell, K. Billis, D. Carvalho-Silva, 626 C. Cummins, P. Clapham, S. Fitzgerald, L. Gil, C. G. GirÃşn, L. Gordon, 627 T. Hourlier, S. E. Hunt, S. H. Janacek, N. Johnson, T. Juettemann, S. Keenan, 628 I. Lavidas, F. J. Martin, T. Maurel, W. McLaren, D. N. Murphy, R. Nag, M. Nuhn, A. Parker, M. Patricio, M. Pignatelli, M. Rahtz, H. S. Riat, D. Sheppard, K. Taylor, A. Thormann, A. Vullo, S. P. Wilder, A. Zadissa, 631 E. Birney, J. Harrow, M. Muffato, E. Perry, M. Ruffier, G. Spudich, S. J. 632 Trevanion, F. Cunningham, B. L. Aken, D. R. Zerbino, and P. Flicek, "Ensembl 633 2016," Nucleic Acids Research, vol. 44, no. D1, pp. D710-D716, 2016. [Online]. 634 Available: + http://dx.doi.org/ 10.1093/nar/gkv 1157635
- 30. V. Haberle, A. R. R. Forrest, Y. Hayashizaki, P. Carninci, and B. Lenhard,
 "CAGEr: precise TSS data retrieval and high-resolution promoterome mining for
 integrative analyses." *Nucleic Acids Research*, vol. 43, no. 8, pp. gkv054–e51, Feb.
 2015.
- 31. H. PagÃÍs, BSgenome: Infrastructure for Biostrings-based genome data packages and support for efficient SNP representation, 2016, r package version 1.42.0.
- 32. H. Thorvaldsdottir, J. T. Robinson, and J. P. Mesirov, "Integrative Genomics Viewer (IGV): high-performance genomics data visualization and exploration,"
 Briefings in Bioinformatics (), vol. 14, no. 2, pp. 178–192, Mar. 2013.
- 33. E. W. E. Sayers, T. T. Barrett, D. A. D. Benson, E. E. Bolton, S. H. S. Bryant, K. K. Canese, V. V. Chetvernin, D. M. D. Church, M. M. Dicuccio, S. S. Federhen, 646 M. M. Feolo, I. M. I. Fingerman, L. Y. L. Geer, W. W. Helmberg, Y. Y. Kapustin, 647 S. S. Krasnov, D. D. Landsman, D. J. D. Lipman, Z. Z. Lu, T. L. T. Madden, 648 T. T. Madej, D. R. D. Maglott, A. A. Marchler-Bauer, V. V. Miller, I. I. Karsch-649 Mizrachi, J. J. Ostell, A. A. Panchenko, L. L. Phan, K. D. K. Pruitt, G. D. G. 650 Schuler, E. E. Sequeira, S. T. S. Sherry, M. M. Shumway, K. K. Sirotkin, D. D. 651 Slotta, A. A. Souvorov, G. G. Starchenko, T. A. T. Tatusova, L. L. Wagner, Y. Y. 652 Wang, W. J. W. Wilbur, E. E. Yaschenko, and J. J. Ye, "Database resources of the 653 National Center for Biotechnology Information." Nucleic Acids Research, vol. 40, no. Database issue, pp. D13-D25, Jan. 2012.
- 34. O. Tange, "Gnu parallel the command-line power tool," ;login: The
 USENIX Magazine, vol. 36, no. 1, pp. 42–47, Feb 2011. [Online]. Available: http://www.gnu.org/s/parallel

⁶⁵⁹ 7 Checklist of Items to be Sent to Volume Editors

660	Here is a checklist of everything the volume editor requires from you:
661	☐ The final LaTeX source files
662	☐ A final PDF file

	Identification of insect regulatory elements	21
A copyright form, signed by paper.	one author on behalf of all of the authors of	the

☐ A readme giving the name and email address of the corresponding author.