

Using RAMPAGE to identify and annotate promoters in insect genomes

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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, shortly thereafter, to reveal widespread transposon-driven alternative promoter usage in *D. melanogaster* [2].

In this chapter we highlight the utility of one TSS profiling method, RAMPAGE (RNA annotation and mapping of promoters for analysis of gene expression), for the precise, quantitative identification of promoters in insect genomes. We demonstrate this using our tools GoRAMPAGE [3] and TSRchitect [4], providing details 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 Introduction

1.1 TSS Profiling Identifies Promoters at Genome-Scale

The promoter, defined in eukaryotes as the genomic region bound by RNA Polymerase II immediately prior to transcription initiation [5], is the site where regulatory signals unite to direct gene expression. The identification of promoter regions is a valuable step for understanding the *cis*-regulatory signals that are present in an organism, and is also important for genome annotation. However, 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

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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).

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 NGS era, CAGE was coupled with massively-parallel sequencing to generate 5'-ends of mRNAs at substantially higher scale. This advance provided more extensive coverage of the expressed transcriptome, and provided increased sensitivity for quantitative measurements *i.e.* measurement of promoter activity.

1.2 Promoter Architecture of *Drosophila melanogaster*

Hoskins and colleagues [1] performed CAGE in *D. melanogaster* as part of the modENCODE consortium, identifying promoters at large-scale and characterizing the promoter architecture of an insect genome for the first time. Hoskins [1] indicated that TSS distributions at *Drosophila* promoters exhibit a range of shapes that can be generally grouped into two major classifications: *peaked* and *broad*. Peaked promoters have a 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]. The authors also showed a strong association between promoter class and motif composition (consistent with previous findings [8, 10]). Peaked promoters were associated with positionally-enriched *cis*-regulatory motifs including TATA, Initiator (Inr) and DPE, while broad promoters contained an enrichment of less-well characterized motifs, including *Ohler6* and *Ohler7* [11]. The existence of two promoter classes appears to be conserved among metazoans, and has been reported (using TSS profiling methodologies) in insects, cladocerans [12], fish [13] and mammals [14, 9].

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: more than 80% of TSSs (81%) 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.

53
 54 The only other insect species for which TSS profiling has been applied is the
 55 Tsetse fly (*Glossina morsitans morsitans*) [15]. Using TSS-seq (specifically Oligo-
 56 capping; for details see [16]), the authors identified 3134 mapping to 1424 genes.
 57 The authors found a preference for CA and AA dinucleotides at the TSS, and
 58 observe the major core promoter elements observed in *Drosophila*: TATA, Inr,
 59 DPE, in addition to MTE (Motif Ten Element). As in *D. melanogaster*, peaked
 60 promoters were more likely to contain TATA and Inr than broad promoters.
 61 While the taxonomic sampling of species for TSS profiling has been limited, the
 62 existing studies are sufficient to provide a general picture of insect promoter ar-
 63 chitecture. A major demarcation between the promoter architecture of insects
 64 and mammals appears to be the large fraction of mammalian promoters found
 65 in CpG islands [15]. CpG island promoters (CPIs) form the largest class of pro-
 66 moter in mammals [17]; by contrast, CPIs are not known to exist as a class in
 67 invertebrates.

68 1.4 Paired-end TSS Profiling with RAMPAGE

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

79
 80 Batut and colleagues [2] generated libraries from total RNA isolated from 36
 81 stages across the life cycle of *D. melanogaster* providing a comprehensive gene
 82 expression and promoter atlas for fruit fly and in the process demonstrating the
 83 utility of RAMPAGE. RAMPAGE is currently being applied as part of the latest
 84 iteration of ENCODE to identify promoters in human, but as of this writing it
 85 has not been applied to any non-*Drosophila* insect model system. In anticipation
 86 of the future application of TSS profiling into other insect model systems here
 87 we provide a documented protocol for the computational processing RAMPAGE
 88 data, using selected libraries from Batut *et al.* [2]. This method will consist of two
 89 parts: first, we will process, filter and align the sequenced RAMPAGE libraries to
 90 the *D. melanogaster* genome. Second, we will identify TSSs and promoters from
 91 the aligned sequences and associate them with coding regions. In closing, we will
 92 consider further applications of this data and discuss the utility of reproducible
 93 workflows in bioinformatic analysis.

94 2 Materials

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

106 2.1 Hardware

- 107 1. x86-64 compatible processors
- 108 2. At least 8GB RAM
- 109 3. 30GB+ hard disk space

110 2.2 Operating System

- 111 – 64 bit Linux (preferred) or Mac OS X (with Command Line Tools from
 112 XCode)

113 2.3 Software

114 Below is a list of the software packages required for this demonstration (*see Note*
 115 **1**).

116 Sequence retrieval

- 117 1. SRA Toolkit [20] (<https://www.ncbi.nlm.nih.gov/sra/docs/toolkitsoft/>)

119 GoRAMPAGE

- 120 1. GoRAMPAGE [3] (<https://github.com/brendelGroup/GoRAMPAGE>)
- 121 2. fastq-multx [21] (<https://github.com/brwnj/fastq-multx>)
- 122 3. FASTX-Toolkit [22] (http://hannonlab.cshl.edu/fastx_toolkit/Index.html)
- 123 4. TagDust2 [23] (<https://sourceforge.net/projects/tagdust/>)
- 124 5. Samtools [24] (<http://www.htslib.org/doc/samtools.html>)
- 125 6. STAR [25] (<https://github.com/alexdobin/STAR>)

126 TSRchitect

- 127 1. R (v. 3.4 and up) [26] (<https://www.r-project.org/>)
- 128 2. Bioconductor (v. 3.5 and up) [27] (<http://bioconductor.org/>)
- 129 3. TSRchitect [4] (<http://bioconductor.org/packages/release/bioc/html/TSRchitect.html>)
- 130 4. Various R package dependencies (see **Methods**)

131 2.4 Online Appendix

132 We created an online appendix to serve as a companion to this chapter, which
 133 contains both scripts and select files to assist you in completing this tutorial.
 134 Please find the repository at https://github.com/rtraborn/MMB_appendix
 135 (see **Note 2**).

136 2.5 Installation of R packages

137 For installation of the software listed above, please follow the instructions pro-
 138 vided by each respective package. Part of our analysis will require the use of R
 139 packages found in the Bioconductor suite [27]. To install Bioconductor, please
 140 type the following from an R console:

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

143 We will use the R package *TSRchitect* to identify promoters from aligned RAM-
 144 PAGE libraries. Prior to running the analysis, it will be necessary to install a
 145 series of prerequisite packages to *TSRchitect* from Bioconductor. Please install
 146 these packages as follows (as before, from an R console):

```
147 source("https://bioconductor.org/biocLite.R")
148 biocLite(c("AnnotationHub", "BiocGenerics", "BiocParallel",
149 "ENCODEExplorer", "GenomicAlignments", "GenomeInfoDb",
150 "GenomicRanges", "IRanges", "methods",
151 "Rsamtools", "rtracklayer", "S4Vectors",
152 "SummarizedExperiment"))
```

153 To install *TSRchitect*, please type the following from an R console:

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

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

```
158 library(TSRchitect) #installing TSRchitect
```

159 3 Methods

160 3.1 Retrieving the RAMPAGE sequence data from NCBI

161 To begin our analysis, we must download the RAMPAGE data to our worksta-
 162 tion. We will utilize tools provided by the SRA Toolkit, which should already
 163 be installed on your machine (see **Materials**). The command *fastq-dump* al-
 164 lows one to directly retrieve data from the GEO database using the appropriate
 165 identifier(s). While there are 36 RAMPAGE libraries in the Batut *et al.* pa-
 166 per, we will select a subset of these to analyze here. We will compare samples

167 from selected embryonic (E01h-E03h) and larval (L1-L3) tissues, representing
 168 the beginning and end of embryonic development. For more information about
 169 the experiment and the available RAMPAGE libraries, please see the following
 170 link: <https://www.ncbi.nlm.nih.gov/Traces/study/?acc=SRP011193>.

171
 172 First, let's proceed with downloading the libraries from early embryonic tissues
 173 (see **See Note 3**). We will make a new folder (entitled "fastq_files/") to
 174 house these files.

```
175 mkdir fastq_files
176 cd fastq_files
177
178 fastq-dump --split-files SRR424683
179 fastq-dump --split-files SRR424684
180 fastq-dump --split-files SRR424685
```

181 We continue by downloading the data from late larval tissues.

```
182 fastq-dump --split-files SRR424707
183 fastq-dump --split-files SRR424708
184 fastq-dump --split-files SRR424709
```

185 Once the download of the aforementioned files are complete, you should see a
 186 total of 12 (6 x 2) separate fastq files in your current working directory:

```
187 ls -l *.fastq | wc -l
188 cd ..
```

189 3.2 Creating symlinks to the files

190 Our workflow expects fastq files that have the format "*.R1/R2.clipped.fq".
 191 Rather than rename them, we can simply create brand new symbolic links (sym-
 192 links) to the files, as follows:

```
193 cd ..
194 mkdir -p output/reads/clipped
195 cd output/reads/clipped
196
197 #embryonic libraries
198 ln -s ../../../../fastq-files/SRR424683_1.fastq E01h.R1.clipped.fq
199 ln -s ../../../../fastq-files/SRR424683_2.fastq E01h.R2.clipped.fq
200 ln -s ../../../../fastq-files/SRR424684_1.fastq E02h.R1.clipped.fq
201 ln -s ../../../../fastq-files/SRR424684_2.fastq E02h.R2.clipped.fq
202 ln -s ../../../../fastq-files/SRR424685_1.fastq E03h.R1.clipped.fq
203 ln -s ../../../../fastq-files/SRR424685_2.fastq E03h.R2.clipped.fq
204
205 #larval libraries
```

```

206 ln -s ../../../../fastq-files/SRR424707_1.fastq L1.R1.clipped.fq
207 ln -s ../../../../fastq-files/SRR424707_2.fastq L1.R2.clipped.fq
208 ln -s ../../../../fastq-files/SRR424708_1.fastq L2.R1.clipped.fq
209 ln -s ../../../../fastq-files/SRR424708_2.fastq L2.R2.clipped.fq
210 ln -s ../../../../fastq-files/SRR424709_1.fastq L3.R1.clipped.fq
211 ln -s ../../../../fastq-files/SRR424709_2.fastq L3.R2.clipped.fq
212
213 cd ../../.. #returning to the output directory

```

214 3.3 Downloading genomic data from *D. melanogaster*

215 Now that we have the fastq files from the RAMPAGE libraries downloaded and
 216 named appropriately, we now must retrieve the genome assembly and rRNA
 217 sequences from *D. melanogaster*. The genome assembly is required for aligning
 218 the RAMPAGE reads, and the rRNA sequences are required to filter out match-
 219 ing reads in the sequenced RAMPAGE libraries, since our sample is intended
 220 to contain only capped RNA transcripts. Please download the rRNA sequences
 221 from the link we provide below. These sequences were retrieved separately from
 222 Genbank at the NCBI database.

223
 224 To retrieve the genome assembly from the ENSEMBL database, please do the
 225 following:

```

226 mkdir genome
227 cd genome
228 wget ftp://ftp.ensembl.org/pub/release-78/fasta/drosophila_melanogaster/dna/Drosophila_m
229 #uncompressing the file
230 gzip -d Drosophila_melanogaster.BDGP5.dna.toplevel.fa.gz
231 cd ..

```

232 Please navigate to the rRNA file "Dmel_rRNA.fasta" found in the Appendix.

```

233 head -n 3
234 >ref|NR_133562.1| Drosophila melanogaster 28S ribosomal RNA (28SrRNA:CR45844), rRNA
235 TTATATACAACCTCAACTCATATGGGACTACCCCTGAATTTAAGCATATTAATTAGGGGAGGAAAAGAA
236 ACTAACAAGGATTTTCTTAGTAGCGGCGAGCGAAAAGAAAACAGTTCAGCACTAAGTCACTTTGTCTATA

```

237 3.4 Filtering and alignment of RAMPAGE reads using 238 GoRAMPAGE

239 At this stage we are ready to commence with the rRNA filtering and alignment
 240 of the RAMPAGE libraries. We will use GoRAMPAGE, a tool we developed,
 241 to perform these tasks in a concerted workflow. GoRAMPAGE runs TagDust
 242 [23] to remove rRNA and low-complexity reads, and uses STAR [25] to align
 243 RAMPAGE (or other paired-end) reads to a given genome assembly.

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

```
249 #vi GoRAMPAGE_script_MMB.sh #updating with a text editor
250 ./GoRAMPAGE_script_MMB.sh
```

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

```
254 less -S errScript
```

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

258 **Inspecting the rRNA filtering results.** To evaluate the results from Step
 259 3 (rRNA filtering), please navigate to the top level of the "output" directory
 260 and open the file "LOGFILES". You'll see the recorded progress of the program
 261 Tagdust and a record of the results. We notice that (for the L3h library) 1046448
 262 of reads (78.1%) were "extracted", meaning that slightly more than 20% of
 263 reads were removed because of matches with ribosomal sequences. The removed
 264 reads from all libraries are found in the "dusted_discard" directory, and the
 265 extracted reads are found in the current directory. Due to their sheer abundance
 266 within cells, ribosomal RNA sequences are an inevitable contaminant within TSS
 267 profiling libraries. For analysis purposes, it is important that these sequences be
 268 removed, which is what has been completed here.
 269 Since this step was conducted appropriately, we can proceed to the next step.

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

282
 283 Now that our RAMPAGE libraries are filtered and aligned, we can commence
 284 with the second half of our analysis.

285 3.5 Promoter identification from aligned RAMPAGE libraries

286 We can now use the prepared alignment files to identify TSSs and promoters from
 287 the selected RAMPAGE libraries. There are currently several tools available
 288 for this purpose. *CAGEr*, developed by Haberle [28], was utilized to perform
 289 TSS identification as part of the FANTOM5 efforts. We will use *TSRchitect* in
 290 this demonstration, since it was specifically designed to analyze paired-end TSS
 291 profiling datasets, and also because it is more flexible with respect to model
 292 system (*i.e.* it does not require a corresponding *BSTGenome* package). The latter
 293 feature will be helpful when analyzing the non-*D. melanogaster* TSS profiling
 294 datasets that we expect to be generated in the near future.

295 **Setting up the Analysis.** *TSRchitect*, the package we'll use for this analy-
 296 sis, is an R package available in the Bioconductor suite of genomics tools [27].
 297 It makes use of existing packages and data structures within this environment,
 298 where available, to identify promoters from sequence alignments. Since you have
 299 already installed *TSRchitect* and its dependencies (see section 2.3), we are set
 300 to proceed.

301
 302 There are two general ways one can choose to run *TSRchitect*. The first is in-
 303 teractively *i.e.* typing the instructions directly into an R console. While this
 304 is a perfectly acceptable way to run analyses using package, for larger jobs
 305 it will likely be more efficient (and likely more reproducible) to run a dedi-
 306 cated R script. We have provided a sample script "`MMB_chapter_TSRchitect.R`"
 307 to make it easier for you to set up an R script. In the section to follow, we
 308 will go through the output of the analysis. For further details on how to use
 309 *TSRchitect*, please see its documentation at its Bioconductor page found here:
 310 <https://www.bioconductor.org/packages/release/bioc/html/TSRchitect.html>.
 311

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

317
 318 Once this is complete, we can run the batch script from the Linux command-line
 319 as follows:

```
320 R CMD BATCH MMB_chapter_TSRchitect.R
321 #assumes variables BAMDIR and DmAnnot have already been set
322 bg #puts this job in the background
```

323 Once the job is underway, you can monitor its progress by looking at the con-
 324 tents of the `.Rout` file (in this case, "`MMB_chapter_TSRchitect.Rout`"). The job
 325 should complete within an hour on most systems.

326

327 **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
 330 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**).

333
 334 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", \
337   inputDir=BAMDIR, inputType="bam", isPairedEnd=TRUE, \
338   sampleNames=c("E1h", "E2h", "E3h", "L1", "L2", "L3"), \
339   replicateIDs=c(1,1,1,2,2,2))
```

340 A critical step in our analysis is identifying TSRs from the aligned TSS data;
 341 to do this we use the function *determineTSR*. We have selected the job to run
 342 on 4 cores in this example (*n.cores*=4). Please enter the number of cores ap-
 343 propriate for your system. Because we want to identify TSRs from every one
 344 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
 346 more 5' RAMPAGE reads will be included within a TSR. Setting *writeTable* to
 347 "TRUE" means that the identified TSRs from each set will be written to the
 348 working directory.

```
349 DmRAMPAGE <- determineTSR(experimentName=DmRAMPAGE, n.cores=4, \
350   tsrSetType="replicates", tssSet="all", tagCountThreshold=25, \
351   clustDist=20, writeTable=TRUE)
```

352 *TSRchitect* can incorporate the tag abundances from each of the samples
 353 and append them to the list of identified TSRs. This is useful for downstream
 354 analysis of differential expression.

```
355 DmRAMPAGE <- addTagCountsToTSR(experimentName=DmRAMPAGE, \
356   tsrSetType="replicates", tsrSet=1, tagCountThreshold=10, \
357   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
 360 TSRs with coding genes. We can specify the maximum distances (both up-
 361 and downstream) between the TSR and the annotation using the arguments
 362 *upstreamDist* and *downstreamDist*.

```
363 DmRAMPAGE <- importAnnotationExternal(experimentName=DmRAMPAGE, \
364   fileType="gff3", annotFile=DmAnnot)
```

```
365  

366 DmRAMPAGE <- addAnnotationToTSR(experimentName=DmRAMPAGE, \
367   tsrSetType="replicates", tsrSet=1, \
368   upstreamDist=1000, downstreamDist=200, feature="gene", \
369   featureColumnID="ID", writeTable=TRUE)
```

Now we have generated a set of identified TSSs, TSRs from all 6 RAMPAGE libraries, and have associated the identified TSRs with annotated genes. Next, we 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 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).

```
#merging the sample data into two groups
DmRAMPAGE <- mergeSampleData(DmRAMPAGE)

# ... identifying TSRs from the merged samples:
DmRAMPAGE <- determineTSR(experimentName=DmRAMPAGE, \
  n.cores=4, tsrSetType="merged", \
  tssSet="all", tagCountThreshold=40, \
  clustDist=20, writeTable=TRUE)
```

Evaluating the results Our analysis using *TSRchitect* is now complete. Your working directory should now contain the following:

- 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)
- TSRs from the combined set of TSSs: TSRsetCombined.tab: (1)

Let's briefly review the files (*see Note 8*). We can quickly obtain the counts on the command line, as follows:

```
wc -l *.tab
8377 TSRset-1.tab
6159 TSRset-2.tab
4814 TSRset-3.tab
17924 TSRset-4.tab
11851 TSRset-5.tab
3242 TSRset-6.tab
13986 TSRsetCombined.tab
7344 TSRsetMerged-1.tab
12126 TSRsetMerged-2.tab
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 differences in library sizes. We detect 7,344 TSRs within the early embryonic samples ("TSRsetMerged-1.tab") and 12,126 TSRs in the late larval samples ("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].

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".

seq	start	end	strand	nTSSs	tsrWidth	shapeIndex	featureID
2L.67043.67044.+	2L	67043	67044	+	270	2	1 NA
2L.74089.74115.+	2L	74089	74115	+	341	27	0.13 NA
2L.94739.94752.+	2L	94739	94752	+	1650	14	0.55 FBgn0031
2L.102386.102386.+	2L	102386	102386	+	284	1	2 FBgn0031

3.6 Summary

The workflow provided here is intended to serve as a useful entry point for the 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 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 profiling technology across a more representative sample of insect diversity will improve our understanding of the positions and general structure *cis*-regulatory regions in this phylum.

3.7 Figures

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 this appendix to your workspace on the command line using git, as follows:

```
git clone https://github.com/rtraborn/MMB_appendix.git
```

The "scripts/" folder in the Appendix 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*).

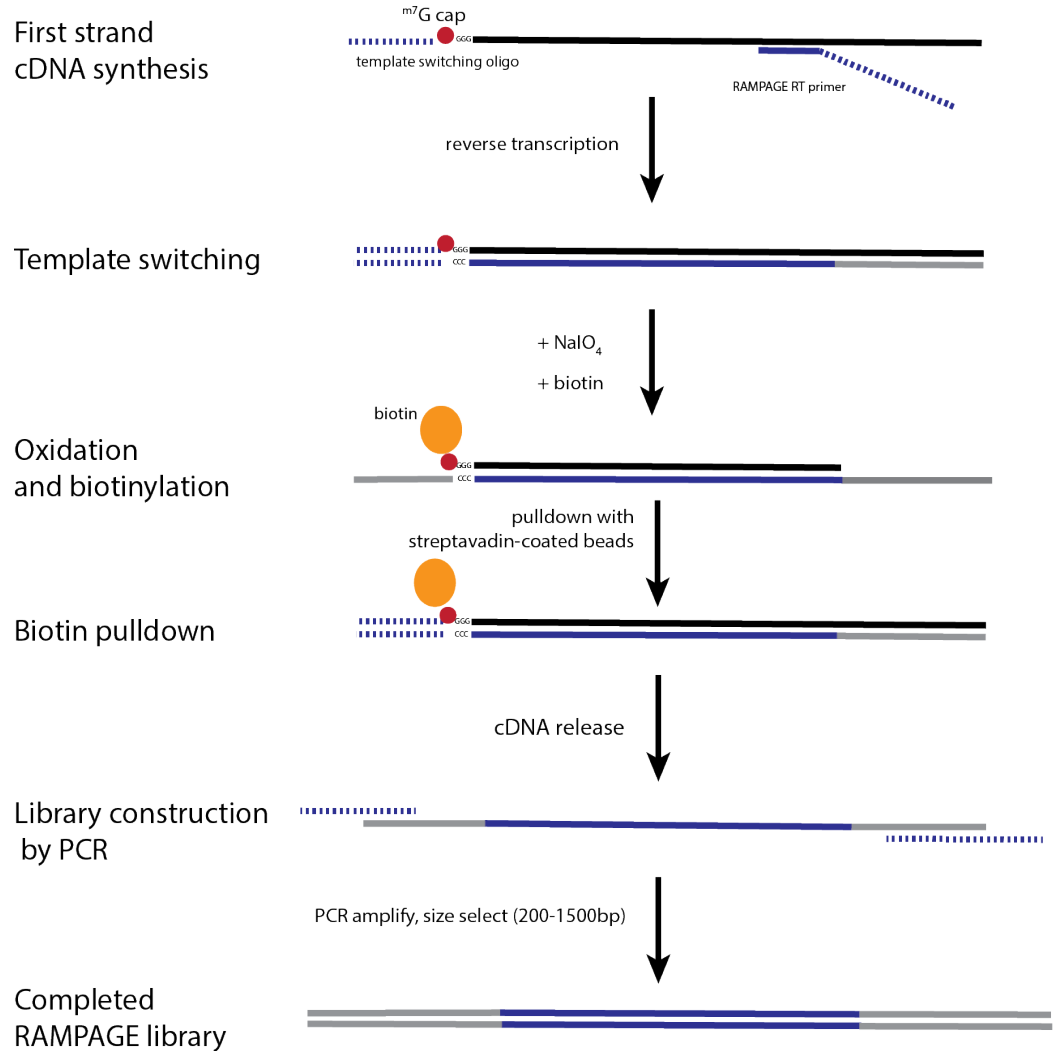


Fig. 1. A brief summary of the RAMPAGE protocol. Starting with high-quality total RNA, first-strand cDNA synthesis is initiated using a cap-bound oligonucleotide and a custom RAMPAGE RT primer, creating a double-stranded DNA-RNA hybrid molecule. Next, the 5'-m7G cap is oxidized, bound with biotin and pulled down with streptavidin-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].

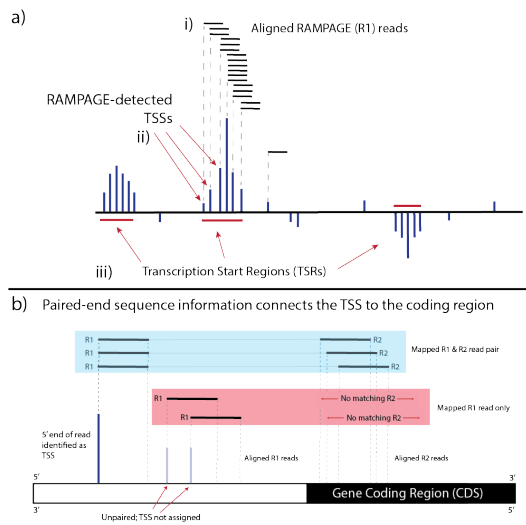


Fig. 2. An overview of promoter identification using RAMPAGE. a) RAMPAGE reads are aligned to the genome. The 5'-most genomic coordinate from each properly-paired R1 read is estimated as a TSS. The abundance of mapped 5'-ends at a given TSS is a measure of its abundance. TSSs above a minimum threshold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

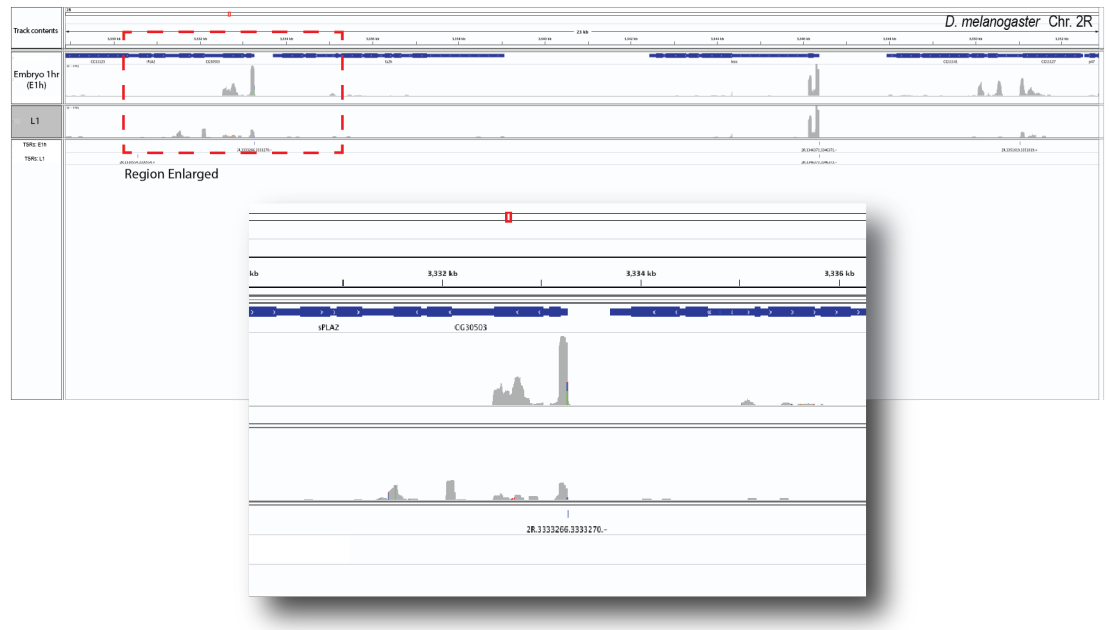


Fig. 3. Test caption for Figure 3

- 450 3. Since these fastq files are paired-end, we use the argument `-split-files` to
451 generate separate files for each read pair.
- 452 4. If you are running this on a cluster with a job scheduler you'll need to add
453 the necessary headers to the top of the script and submit the job in the
454 appropriate manner.
- 455 5. For parallel execution, GoRAMPAGE uses the Linux package *GNU parallel*
456 [29]. Please see the GoRAMPAGE documentation for more information.
- 457 6. To do this, please edit the batch script `TSRchitect_script_MMB.R` with a
458 text editor of your choice.
- 459 7. Because the samples provided derive from related developmental stages, we
460 will merge them for annotation purposes using the argument `replicateIDs`,
461 (though it must be emphasized that they are not replicates).
- 462 8. All of *TSRchitect*'s output files are labeled according to the order that they
463 are loaded onto the *tssObject*. For example, *TSSset-1.txt* corresponds to the
464 first RAMPAGE dataset (in our case E1h), and *TSSset-2.txt* corresponds to
465 the second RAMAPGE dataset (for this example E2h), and so on. You can
466 check which datasets are loaded on the *tssObject* by simply entering it on an
467 R console. Please see the *TSRchitect* documentation for more information.

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Disclosure Declaration

The authors declare that they have no competing interests.

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6 Checklist of Items to be Sent to Volume Editors

Here is a checklist of everything the volume editor requires from you:

- ☐ The final L^AT_EX source files
- ☐ A final PDF file
- ☐ A copyright form, signed by one author on behalf of all of the authors of the paper.
- ☐ A readme giving the name and email address of the corresponding author.