Pooling RNA-seq and Assembling Models

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1 Introduction

Pooling RNA-seq and Assembling Models (PRAM) is an R package that utilizes multiple RNA-seq datasets to predict transcript models. The workflow of PRAM contains four steps. Figure 1 shows each step with function name and associated key parameters. In the later sections of this vignette, we will describe each function in details.

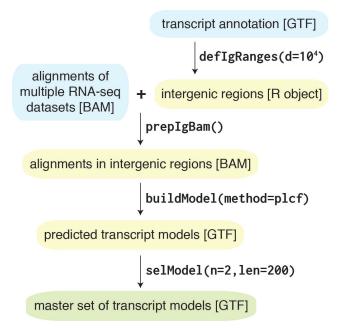


Figure 1: PRAM workflow

2 Installation

Use the following R command on Linux or macOS

```
devtools::install_github('pliu55/pram')
```

3 Quick start

PRAM provides a function runPRAM() to let you run through the whole workflow.

For a given gene annotation and RNA-seq alignments, you can predict transcript models in intergenic genomic regions:

```
runPRAM(in_gtf, in_bamv, out_gtf)
```

- in_gtf: an input GTF file defining genomic coordinates of existing genes. Required to have an attribute of **gene_id** in the ninth column.
- in_bamv: a vector of input BAM file(s) containing RNA-seq alignments. Currently, PRAM only supports strand-specific paired-end RNA-seq with the first mate on the right-most of transcript coordinate, i.e., 'fr-firststrand' by Cufflinks definition.
- out_gtf: an output GTF file of predicted transcript models

3.1 Examples

PRAM has included input examples files in its extdata/demo/ folder. The table below provides a quick summary of all the example files.

Table 1: runPRAM()'s input example files

input argument	file name(s)		
in_gtf	in.gtf		
in_bamv	SZP.bam, TLC.bam		

You can access example files by system.file() in R, e.g. for the argument in_gtf , you can access its example file by

```
system.file('extdata/demo/in.gtf', package='pram')
```

Below shows usage of runPRAM() with example input files:

4 Define intergenic genomic ranges: defigRanges()

To predict intergenic transcripts, we must first define intergenic regions by defIgRanges(). This function requires a GTF file containing known gene annotation supplied for its in_gtf argument. This GTF file should contain an attribue of **gene_id** in its ninth column. We provided an example input GTF file in PRAM package: extdata/gtf/defIGRanges_in.gtf.

In addition to gene annotation, defIgRanges() also requires user to provide chromosome sizes so that it would know the maximum genomic ranges. You can provide one of the following arguments:

- chromgrs: a GRanges object, or
- genome: a genome name, currently supported ones are: hg19, hg38, mm9, and mm10, or
- fchromsize: a UCSC genome browser-style size file, e.g. hg19

By default, defIgRanges() will define intergenic ranges as regions 10 kb away from any known genes. You can change it by the radius argument.

4.1 Example

5 Prepare input RNA-seq alignments: prepigBam()

Once intergenic regions were defined, prepIgBam() will extract corresponding RNA-seq alignments from input BAM files. In this way, transcript models predicted at later stage will solely from intergenic regions. Also, with fewer RNA-seq alignments, model prediction will run faster.

Three input arguments are required by prepIgBam():

- finbam: an input RNA-seq BAM file sorted by genomic coordinate. Currently, we only support strand-specific paired-end RNA-seq data with the first mate on the right-most of transcript coordinate, i.e. 'fr-firststrand' by Cufflinks's definition.
- iggrs: a GRanges object to define intergenic regions.
- foutbam: an output BAM file.

5.1 Example

6 Build transcript models: buildModel()

buildModel() predict transcript models from RNA-seq BAM file(s). This function requires two arguments:

- in_bamv: a vector of input BAM file(s)
- out_gtf: an output GTF file containing predicted transcript models

6.1 Transcript prediction methods

yes

plst

buildModel() has implemented seven transcript prediction methods. You can specify it by the method argument with one of the keywords: plcf, plst, cfmg, cftc, stmg, cf, and st. The first five denote meta-assembly methods that utilize multiple RNA-seq datasets to predict a single set of transcript models. The last two represent methods that predict transcript models from a single RNA-seq dataset.

The table below compares prediction steps for these seven methods. By default, buildModel() uses **plcf** to predict transcript models.

StringTie

no

method	meta- assembly	preparing RNA-seq input	building transcripts	assembling transcripts
plcf	yes	pooling alignments	Cufflinks	no

Table 2: Prediction steps of the seven buildModel() methods

pooling alignments

method	meta- assembly	preparing RNA-seq input	building transcripts	assembling transcripts
cfmg	yes	no	Cufflinks	Cuffmerge
cftc	yes	no	Cufflinks	TACO
stmg	yes	no	StringTie	StringTie-merge
cf	no	no	Cufflinks	no
st	no	no	StringTie	no

6.2 Required external software

Depending on your specified prediction method, buildModel() requires external software: Cufflinks, StringTie and/or TACO, to build and/or assemble transcript models. You can either specify the software location using the cufflinks, stringtie, and taco arguments in buildModel(), or simply leave these three arugments undefined and let PRAM download them for you automatically. The table below summarized software versions buildModel() would download when required software was not specified. Please note that, for macOS, pre-compiled Cufflinks binary versions 2.2.1 and 2.2.0 appear to have an issue on processing BAM files, therefore we recommend to use version 2.1.1 instead.

Table 3: buildModel()-required software and recommended version

software	Linux binary	macOS binary	required by	
Cufflinks, Cuffmerge	v2.2.1	v2.1.1	plcf, cfmg, cftc, and cf	
StringTie, StringTie-merge	v1.3.3b	v1.3.3b	plst, stmg, and st	
TACO	v0.7.0	v0.7.0	cftc	

6.3 Example

7 Select transcript models: **selModel()**

Once transcript models were built, you may want to select a subset of them by their genomic features. selModel() was developed for this purpose. It allows you to select transcript models by their total number of exons and total length of exons and introns.

selModel() requires two arguments:

- fin_gtf:input GTF file containing to-be-selected transcript models. This file is required to have **transcript_id** attribute in the ninth column.
- fout_gtf: output GTF file containing selected transcript models.

By default: selModel() will select transcript models with ≥ 2 exons and ≥ 200 bp total length of exons and introns. You can change the default using the min_nexon and min_tr_len arguments.

7.1 Example

```
fin_gtf = system.file('extdata/gtf/selModel_in.gtf', package='pram')
fout_gtf = tempfile(fileext='.gtf')
selModel(fin_gtf, fout_gtf)
```