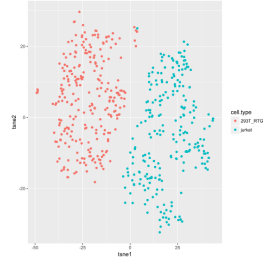


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## 🌐 Instructional tutorial for using *demuxlet*

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External link: <https://github.com/statgen/demuxlet/tree/master/tutorial>



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## Abstract

### Genetic multiplexing of barcoded single cell RNA-seq

#### Introduction

**demuxlet** is a software tool to deconvolute sample identity and identify multiplets when multiple samples are pooled by barcoded single cell sequencing. **demuxlet** takes (1) a SAM/BAM/CRAM file produced by the standard 10x sequencing platform, or any other barcoded single cell RNA-seq (with proper `--tag-UMI` and `--tag-group`) options (2) a VCF/BCF file containing the genotype (GT), posterior probability (GP), or genotype likelihood (GL) to assign each barcode to a specific sample (or a pair of samples) in the VCF file.

#### Tutorial for demuxlet

This tutorial provides streamlined instructions for using the tool **demuxlet**. For a more detailed description of all of the options available to use with **demuxlet**, please refer to the README (attached in .zip file). **demuxlet** has several dependencies and this tutorial provides steps to use a docker instance to minimize compatibility issues and an alternative of installing demuxlet from source.

**demuxlet** is a software tool to deconvolute sample identity and identify multiplets when multiple samples are pooled by barcoded single cell sequencing. **demuxlet** requires the following input files:

1. a SAM/BAM/CRAM file produced by the standard 10x sequencing platform, or any other barcoded single cell RNA-seq (with proper `--tag-UMI` and `--tag-group`) options.
2. a VCF/BCF file containing the genotype (GT), posterior probability (GP), or genotype likelihood (GL) to assign each barcode to a specific sample (or a pair of samples) in the VCF file.

## Attachments



[tutorial.zip](#)

2.7MB



## Guidelines

### Additional resources

The README for demuxlet is available [here](#).

If you have questions about using demuxlet or suggestions for future releases, please contact [jimmie.ye@ucsf.edu](mailto:jimmie.ye@ucsf.edu).

### Source of 293T and Jurkat VCF file.

1. 293T VCF Source website: <http://hek293genome.org/v2/data.php> Source file: [http://bioinformatics.psb.ugent.be/downloads/genomeview/hek293/SNP/293T\\_RTG.vcf.gz](http://bioinformatics.psb.ugent.be/downloads/genomeview/hek293/SNP/293T_RTG.vcf.gz)
2. Jurkat VCF Source website: <https://zenodo.org/record/400615#.WYlh7lQrLlV> Source file: [https://zenodo.org/record/400615/files/jurkat\\_final\\_variant\\_calls.tar.gz](https://zenodo.org/record/400615/files/jurkat_final_variant_calls.tar.gz)
3. 293T:jurkat VCF file generation We used the CrossMap tool to liftover the 293T vcf file from hg18 to hg19. The tetraploid genotype for the Jurkat vcf was collapsed to a diploid genotype before being merged with the 293T vcf file and the resulting file was filtered to contain only the exon positions.

## Installing demuxlet

### 1. MAC NOTES

You will need to install some additional packages on a Mac.

First, you will need XCode CommandLineTools

Second, it's good to install homebrew:

```
/usr/bin/ruby -e "$(curl -fsSL  
https://raw.githubusercontent.com/Homebrew/install/master/install)"
```

Then, we need to install autoconf, automake, and libtool

```
$ brew install autoconf automake libtool
```

We will also need a new clang since Apple's clang doesn't support OpenMPI



```
$ brew install llvm
$ export CC=/usr/local/Cellar/llvm/4.0.0_1/bin/clang
$ export CXX=/usr/local/Cellar/llvm/4.0.0_1/bin/clang++
$ export LDFLAGS=-L/usr/local/Cellar/llvm/4.0.0_1/lib/
```

The CRAM format may use LZMA2 compression, which is implemented in HTSlib by using compression routines from liblzma <http://tukaani.org/xz/>.

Building HTSlib requires liblzma development files to be installed on the build machine; you may need to ensure a package such as liblzma-dev (on Debian or Ubuntu Linux), xz-devel (on RPM-based Linux distributions or Cygwin), or xz (via Homebrew on macOS) is installed; or build XZ Utils from source. Assuming we are on a Mac:

```
$ brew install xz
```

## 2. Install htslib

```
$ git clone https://github.com/samtools/htslib.git
$ cd path/to/htslib
$ autoheader
$ autoconf
$ ./configure # optional, --prefix=/path/file
$ make
$ make install # optional, DESTDIR=/path/file
```

## 3. Install demuxlet

demuxlet and htslib should be installed in same directory

```
$ git clone https://github.com/statgen/demuxlet.git
$ cd /path/to/demuxlet
$ mkdir m4
$ autoreconf -vfi
$ ./configure # optional, --prefix=/path/file
$ make
$ make install # optional, DESTDIR=/path/file
```

## Tips for running



- Set **--alpha 0 --alpha 0.5**, which assumes the expected proportion of 50% genetic mixture from two individuals, to get better estimates of doublets.
- Set **--group-list** to a list of barcodes (i.e. barcodes.tsv from 10X) to speed things up and only get demultiplexing for cells called by other methods
- To reproduce the results presented in Figure 2 of the demuxlet paper, please go to: [https://github.com/yelabucsf/demuxlet\\_paper\\_code/tree/master/fig2](https://github.com/yelabucsf/demuxlet_paper_code/tree/master/fig2) to download the vcf and the outputs of demuxlet.

## Using demuxlet

demuxlet uses a self-documentation utility. You can run each utility with -man or -help option to see the command line usages.

```
$ ./demuxlet          (for short usage)
$ ./demuxlet -help    (for detailed usage)
```

The detailed usage is also pasted below.

## Options for input SAM/BAM/CRAM

--sam [STR: ] : Input SAM/BAM/CRAM file. Must be sorted by coordinates and indexed

--tag-group [STR: CB] : Tag representing readgroup or cell barcodes, in the case to partition the BAM file into multiple groups. For 10x genomics, use CB

--tag-UMI [STR: UB] : Tag representing UMIs. For 10x genomics, use UB

## Options for input VCF/BCF

--vcf [STR: ] : Input VCF/BCF file, containing the individual genotypes (GT), posterior probability (GP), or genotype likelihood (PL)

--field [STR: GP] : FORMAT field to extract the genotype, likelihood, or posterior from

--geno-error [FLT: 0.01] : Genotype error rate (must be used with --field GT)

--min-mac [INT: 1] : Minimum minor allele frequency --min-callrate [FLT: 0.50] : Minimum call rate

--sm [V\_STR: ] : List of sample IDs to compare to (default: use all)

--sm-list [STR: ] : File containing the list of sample IDs to compare

## Output Options

--out [STR: ] : Output file prefix

--alpha [V\_FLT: ] : Grid of alpha to search for (default is 0.1, 0.2, 0.3, 0.4, 0.5)

--write-pair [FLG: OFF] : Writing the (HUGE) pair file

--doublet-prior [FLT: 0.50] : Prior of doublet

--sam-verbose [INT: 1000000] : Verbose message frequency for SAM/BAM/CRAM

--vcf-verbose [INT: 10000] : Verbose message frequency for VCF/BCF

## Read filtering Options

--cap-BQ [INT: 40] : Maximum base quality (higher BQ will be capped)

--min-BQ [INT: 13] : Minimum base quality to consider (lower BQ will be skipped)

--min-MQ [INT: 20] : Minimum mapping quality to consider (lower MQ will be ignored)

--min-TD [INT: 0] : Minimum distance to the tail (lower will be ignored)

--excl-flag [INT: 3844] : SAM/BAM FLAGS to be excluded

## Cell/droplet filtering options

```
--group-list      [STR: ]                : List of tag readgroup/cell barcode to
consider in this run. All other barcodes will be ignored. This is useful for
parallelized run
--min-total       [INT: 0]                : Minimum number of total reads for a
droplet/cell to be considered
--min-uniq        [INT: 0]                : Minimum number of unique reads (determined by
UMI/SNP pair) for a droplet/cell to be considered
--min-snp         [INT: 0]                : Minimum number of SNPs with coverage for a
droplet/cell to be considered
```

## Interpretation of output files

**demuxlet** generates multiple output file, such as **[prefix].best**, **[prefix].sing**, **[prefix].sing2**, and optionally **[prefix].pair** (with **--write-pair** argument). Each file contains the following information

- The **[prefix].best** file contains the best guess of the sample identity, with detailed statistics to reach to the best guess
- The **[prefix].sing** file contains the statistics for matching each cell with each possible sample.
- The **[prefix].sing2** file contains the statistics similar information to the previous one, but generated for sanity checking of the **[prefix].pair** results.
- The **[prefix].pair** file contains the statistics for matching each cell with each possible configuration of doublet.

The **[prefix].best** file contains the following 22 columns.

1. BARCODE - Cell barcode for the cell that is being assigned in this row
2. RD.TOTL - The total number of reads overlapping with variant sites for each droplet.
3. RD.PASS - The total number of reads that passed the quality threshold, such as mapping quality, base quality.
4. RD.UNIQ - The total number of UMIs that passed the quality threshold. If a UMI is observed in a single variant multiple times, it won't be counted more. If a UMI is observed across multiple variants, it will be counted as different.
5. N.SNP - The total number of variants overlapping with any read in the droplet.
6. BEST - The best assignment for sample ID.
  - For singlets, SNG-
  - For doublets, DBL---
  - For ambiguous droplets, , AMB---<doublet ID1/ID2>)
7. SNG.1ST - The best singlet assignment for sample ID
8. SNG.LLK1 - The log(likelihood that the ID from SNG.1ST is the correct assignment)
9. SNG.2ND - The next best singlet assignment for sample ID
10. SNG.LLK2 - The log(likelihood that the ID from SNG.2ND is the correct assignment)
11. SNG.LLK0 - The log-likelihood from allele frequencies only
12. DBL.1ST - The sample ID that is most likely included if the assignment is a doublet





13. DBL.2ND - The sample ID that is next most likely included if the assignment is a doublet
14. ALPHA - % Mixture Proportion
15. LLK12 - The log(likelihood that the ID is a doublet)
16. LLK1 - The log(likelihood that the ID from DBL.1ST is the correct singlet assignment)
17. LLK2 - The log(likelihood that the ID from DBL.2ND is the correct singlet assignment)
18. LLK10 - The log(likelihood that the ID from DBL.1ST is one of the doublet, and the other doublet identity is calculated from allele frequencies only)
19. LLK20 - The log(likelihood that the ID from DBL.2ND is one of the doublet, and the other doublet identity is calculated from allele frequencies only)
20. LLK00 - The log(likelihood that the droplet is doublet, but both identities are calculated from allele frequencies only)
21. PRB.DBL - Posterior probability of the doublet assignment
22. PRB.SNG1 - Posterior probability of the singlet assignment when excluding all possible doublets

## Troubleshooting

### Before start

Add bam file name for **\$bam** and vcf file name for **\$vcf**. Use **<(zcat \$vcf)** or **<(gzcat \$vcf)** if vcf file is compressed.

The options for **--field** are individual genotypes (GT), posterior probability (GP), or genotype likelihood (PL). If using GT **--field** option for, you must include **--geno-error**, which is the genotype error rate.

### demuxlet output

The demuxlet software produces 3 output files. Among those, most important are:

1. **[prefix].best** The .best file contains the assignments of the best sample identity (singlet: **SNG-**; doublet: **DBL-**; ambiguous: **AMB-< >**) in the **BEST** column for each cell barcode identified in the **BARCODE** column along with details of the statistics used to determine the best identity.
2. The **[prefix].single** file contains the statistics for matching each cell with each possible sample

For complete descriptions of the generated files and columns in the output, please see the [Guidelines](#).



1 Please select between the two following options:

1. Installing from Docker
2. Installing from source

## STEP CASE

### Installing from Docker 6 steps

Select this option for docker installation.

#### 2 Installing docker

First, get docker for whatever platform you feel comfortable with:

<https://www.docker.com/get-started>

#### 3 Running demuxlet from docker

We have created a docker container at **yimmieg/demuxlet** to run demuxlet through docker.

You can run it with

```
$ docker run -v path/to/tutorial:/data yimmieg/demuxlet --sam /data/$bam --vcf /data/$vcf --field $(GT or GP or PL) --out /data/$filename
```

## Analyzing the sample dataset

#### 4 Create directory and download datasets

Now, let's first download the data we need. We are now providing a one-stop-shop to download all of the data you need for the tutorial:

<https://ucsf.box.com/s/vg1bycvsjgyg63gkqspuq5rxzjl6k>. After downloading and unzipping (it's BIG), you should have a directory called **demuxlet.tutorial**.

#### 5 Run demuxlet

```
$ cd demuxlet.tutorial
$ docker run -v ./:/data yimmieg/demuxlet --sam
/data/jurkat_293t_downsampled_n500_full_bam.bam --vcf
/data/jurkat_293t_exons_only.vcf --field GT --out
data/jurkat_293t_demuxlet
```

## 6 Compare the called genotypes vs transcriptome data

In this analysis, we will use R to produce a t-SNE (t-Distributed Stochastic Neighbor Embedding) plot of the cells from the 293T:Jurkat 10x experiment with the cells colored by the assignments from the demuxlet pipeline.

```
library(ggplot2);
library(data.table);

## let's read in the barcodes
tsne <- fread("analysis_csv/tsne/projection.csv");
demuxlet <- fread("jurkat_293t_demuxlet.best");

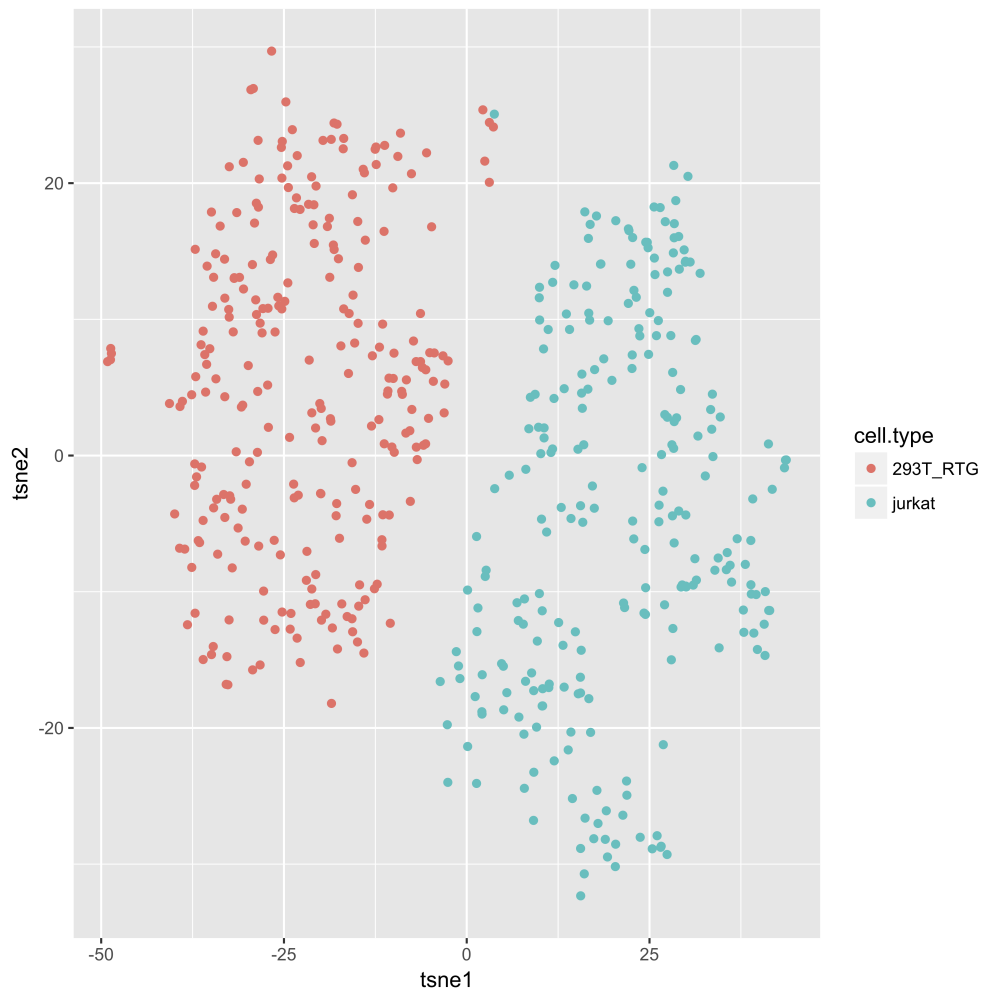
## let's filter for the barcodes that we sampled

df <- data.frame(tsne1=tsne$"TSNE-1"
[na.omit(match(demuxlet$BARCODE, tsne$Barcode))], tsne2=tsne$"TSNE-
2"[na.omit(match(demuxlet$BARCODE, tsne$Barcode))],
doublet=sapply(demuxlet$BEST, function(x){strsplit(x, "-")[[1]]
[[1]]}), cell.type=sapply(demuxlet$BEST, function(x){strsplit(x, "-
")[[1]][[2]]}))

ggplot(aes(tsne1, tsne2, color=cell.type), data=df)+geom_point()
```

## Expected result

After this, you should get the following image:



7 Let's also take a look at the doublets.

```
ggplot(aes(tsne1, tsne2, color=doublet), data=df)+geom_point()
```

## Expected result

After that, you should get the following image:

