How to analyse single cell SLAMseq data (as of 23.03.2022)

--> global MT tag version

0. setting up:

make sure to have a conda environment (if desired, python 3.8 tested) and the following packages in working order:

```
pysam
re
pdb
sys
os
subprocess
string
pandas
numpy
```

Only pysam needs to be installed, the others should come with your python installation.

pandas and numpy are not strictly necessary for this particular workflow, but do appear in various analysis scripts that may or may not be used later on. They can be skipped here if desired.

have a vaguely recent version of samtools installed (via conda or otherwise, v. 1.14 tested)

1. map with cellranger similar to this (cellranger 6.1.1 tested):

```
<location/of/your/cellranger>/cellranger count
--id=<name for the output folder>
--fastqs=/data/junker/runs/<folder with your fastqs>
--transcriptome=<your transcriptome here>
--sample=<sample name from demultiplexing>
--localcores=<20 to 40>
```

2. extract actual cells (as detected by cellranger) if desired. Can be skipped if cellranger is not to be trusted on your data

```
enter the /outs/filtered_feature_bc_matrix/ directory and unzip barcodes.tsv.gz
```

run

```
python <script location>/1_extract_actual_cells_v2.py possorted_genome_bam.bam \
/outs/filtered_feature_bc_matrix/barcodes.tsv N
```

3. add information about the reference genome for each mismatch using samtools:

```
samtools calmd possorted_genome_bam_actual_cells.bam <path/to/genome>/genome.fa \
   | samtools view -Sb | samtools sort > possorted_genome_bam_actual_cells_MD.bam
```

4. aggregate information on mutations in a custom tag:

```
python <script location>2_create_MT_tag_v4_global_toStdout.py possorted_genome_bam_actual_cells_MD.bam
```

catch the script's output as desired, instructions will appear automatically if the output is not piped.

This guide assumes the output of Step 4 to be called possorted_genome_bam_actual_cells_MD_MTglob.bam

index the output

```
samtools index possorted_genome_bam_actual_cells_MD_MTglob.bam
```

5. separate labelled and unlabelled reads:

```
python \
  <script location>3_separate_labeled_unlabeled_reads_v5_2mutations_MTglob_enhancedUmiStats_PrimOnly.py \
   possorted_genome_bam_actual_cells_MD_MTglob.bam <mutation sequencing quality> <number of mutations per UMI>
```

This will generate 3 .bam files, one that has all UMIs that satisfy the 'labelled' criteria, one with entirely unlabelled UMIs and the third one with unclear cases that appear when more that 1 mutation per UMI is required in order to determine it to be labelled.

6. calculate bulk mutation rates:

run

```
python <script location>/4_count_mutations_SE_MTglob_v6.py \
possorted_genome_bam_actual_cells_MD_MTglob.bam <mutation sequencing quality>
```

This generates one file that has total nucleotice counts (based on the reference in case of substitutions) and one that has total mutation counts. Use to plot as desired.

7. generate labelled / unlabelled count matrices

generate the count matrices directly from the .bam files based on a script based on one that was supplied with STARsolo 2.7.10a (STAR_2.7.10a/extras/scripts/soloCountMatrixFromBAM.awk)

run

```
python <script location>/5_count-matrix_from_bam_v1.py \
possorted_genome_bam_actual_cells_MD_MTglob_(un)labeled_<...> <desired matrix output folder>
```

zip all three files in <desired matrix output folder> by e.g. entering it and gzip *

The resulting matrices are in MatrixMarket matrix coordinate integer general format, similar to the cellranger and STARsolo output

In case of Seurat analysis: feed said <desired matrix output folder> to seurat as a 10x directory like this:

```
project.data <- Read10X(data.dir = "<desired matrix output folder>", gene.column=1)
```

and proceed as normal