

Look more into fixed versus adaptive thresholds for QC:

<http://bioconductor.org/books/3.13/OSCA.basic/quality-control.html#identifying-low-quality-cells>

<https://doi.org/10.1186/s13059-020-02136-7> -> practical recommendations, also interesting plots in supplementary

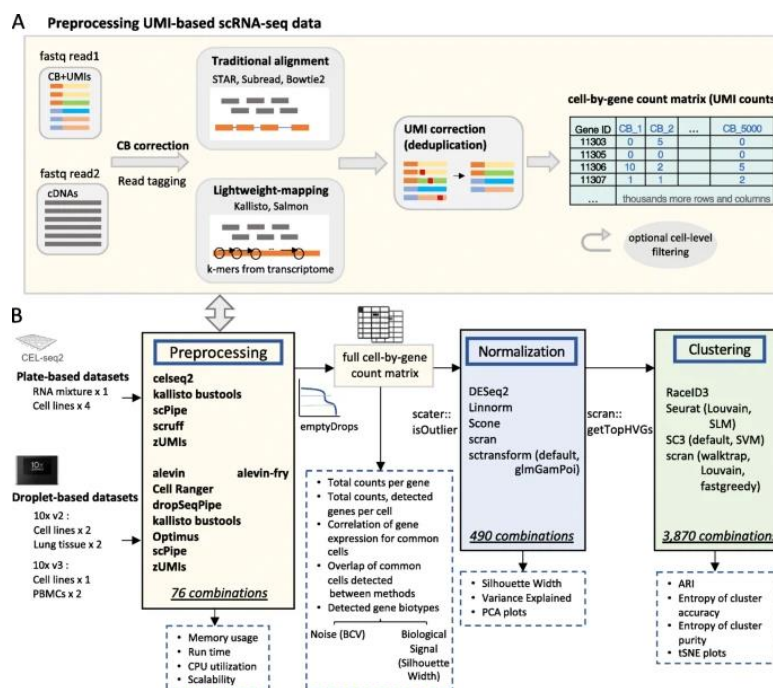
Performance Assessment and Selection of Normalization Procedures for Single-Cell RNA-Seq:

<https://www.bioconductor.org/packages/release/bioc/vignettes/scone/inst/doc/sconeTutorial.html>

Current best practices in single-cell RNA-seq analysis: a tutorial -> There is no consensus on scaling genes to 0 mean and unit variance. We prefer not to scale gene expression

Diagnostic plots for filtering, normalization, feature selection:

Try a simply normalization technique like log1pPF and assess normalization result by visualizing it on a UMAP with respect to total counts and highly expressed genes in your dataset.



<https://doi.org/10.1186/s13059-021-02552-3>

[https://www.sc-best-practices.org/preprocessing\\_visualization/normalization.html](https://www.sc-best-practices.org/preprocessing_visualization/normalization.html)

“We cannot treat all genes the same”: [https://hbctraining.github.io/scRNA-seq\\_online/lessons/06\\_SC\\_SCT\\_normalization.html](https://hbctraining.github.io/scRNA-seq_online/lessons/06_SC_SCT_normalization.html)

[https://htmlpreview.github.io/?https://github.com/satijalab/sctransform/blob/supplement/variante\\_stabilizing\\_transformation.html](https://htmlpreview.github.io/?https://github.com/satijalab/sctransform/blob/supplement/variante_stabilizing_transformation.html)

PCA: [https://hbctraining.github.io/scRNA-seq\\_online/lessons/05\\_theory\\_of\\_PCA.html](https://hbctraining.github.io/scRNA-seq_online/lessons/05_theory_of_PCA.html)

Normalization and variance stabilization of single-cell RNA-seq data using regularized negative binomial regression, <https://doi.org/10.1186/s13059-019-1874-1>,

<https://www.biorxiv.org/content/10.1101/2021.07.07.451498v1>

“For reference, we first apply the standard Seurat workflow, with log-normalization”

“For comparison, we now apply sctransform normalization” # Note that this single command replaces NormalizeData, ScaleData, and # FindVariableFeatures.

[https://htmlpreview.github.io/?https://github.com/satijalab/sctransform/blob/supp\\_html/supplement/seurat.html](https://htmlpreview.github.io/?https://github.com/satijalab/sctransform/blob/supp_html/supplement/seurat.html)

Implementation: [https://satijalab.org/seurat/articles/sctransform\\_v2\\_vignette.html](https://satijalab.org/seurat/articles/sctransform_v2_vignette.html)

pipeComp, a general framework for the evaluation of computational pipelines, reveals performant single cell RNA-seq preprocessing tools, <https://doi.org/10.1186/s13059-020-02136-7>

-> **We might want to keep more features if we can go on PSB cluster (up to 4000)**

[https://www.sc-best-practices.org/cellular\\_structure/annotation.html](https://www.sc-best-practices.org/cellular_structure/annotation.html)

Manual annotation versus other methods:

For manual annotation: get marker genes from <https://training.galaxyproject.org/training-material/topics/single-cell/tutorials/scrna-plant/tutorial.html>

Tutorial: [https://hbctraining.github.io/scRNA-seq\\_online/lessons/09\\_merged\\_SC\\_marker\\_identification.html](https://hbctraining.github.io/scRNA-seq_online/lessons/09_merged_SC_marker_identification.html)

<https://www.sciencedirect.com/science/article/pii/S2001037021000192>

<http://bioconductor.org/books/release/SingleRBook/introduction.html>

<https://github.com/Irrationone/cellassign>

<https://docs.scvi-tools.org/en/stable/tutorials/index.html>

Integration: [https://satijalab.org/seurat/articles/sctransform\\_v2\\_vignette.html](https://satijalab.org/seurat/articles/sctransform_v2_vignette.html)

From Louvain to Leiden: guaranteeing well-connected communities

Bias, robustness and scalability in single-cell differential expression analysis