## 0 Introduction

Doublets are a characteristic error source in droplet-based single-cell sequencing data where two cells are encapsulated in the same oil emulsion and are tagged with the same cell barcode. Across-type doublets manifest as fictitious phenotypes that can be incorrectly interpreted as novel cell types. We present a novel, fast, unsupervised classifier to detect across-type doublets in single-cell RNA-sequencing data that operates on a count matrix and imposes no experimental constraints. This classifier leverages the creation of in silico synthetic doublets to determine which cells in the input count matrix have gene expression that is best explained by the combination of distinct cell types in the matrix. In the next section, we invite you to explore the pseudocode of our method. A bioRxiv submission is currently in the works.

## 1 Algorithms

**Algorithm 1.1** Downsampling of two parent gene-vectors. *cell*1 and *cell*2 are gene-vectors represented as arrays of gene indices. Each element of the array is an index corresponding to one transcript of that gene in the gene-vector. If a gene-vector has, for example, 15 counts of a particular gene, the corresponding index will be present 15 times in the array.

```
1: function DOWNSAMPLECELLPAIR(cell1, cell2)
2: newsize \leftarrow \max(\operatorname{length}(cell1), \operatorname{length}(cell2))
3: summed \leftarrow \operatorname{append}(cell1, cell2)
4: shuffled \leftarrow \operatorname{RandomPermutation}(summed) \triangleright Shuffle position of elements.
5: synthetic \leftarrow shuffled[: newsize]
6: \mathbf{return}\ synthetic
7: \mathbf{end}\ \mathbf{function}
```

**Algorithm 1.2** Cluster p-value Assignment via Hypergeometric Test. N is the number of cells in the augmented dataset. K is the number of synthetic doublets in the augmented dataset. clusters is the set of clusters for the augmented dataset, where each cluster contains its member cells.

```
1: function AssignP-Values(N, K, clusters)
        P \leftarrow \text{empty } N \text{ length array}
 2:
        for c in clusters do
 3:
             n \leftarrow |c|
 4:
             k \leftarrow number of synthetic doublets in c
 5:
             p \leftarrow \text{hypergeom.cdf}(N, K, n, k)
 6:
             for cell in c do
 7:
                  P[\operatorname{IndexOf}(cell)] \leftarrow p
 8:
             end for
 9:
        end for
10:
        return P
12: end function
```

**Algorithm 1.3** Classify cells as doublets or singlets. counts is an N by D count matrix, where N is the number of cells and D is the number of genes.

**Precondition:** ITERS and BOOSTRATE have been set.

```
1: function DoubletDetection(counts)
        N, D \leftarrow \text{shape}(counts)
                                                                                         \triangleright Dimensions of counts.
 3:
        doublet \leftarrow \text{empty } ITERS \text{ by } N \text{ boolean matrix}
        for i = 1, \dots, ITERS do
 4:
            raw\_synths \leftarrow \text{CreateDoublets}(counts, BOOSTRATE)
 5:
            augmented \leftarrow NormalizeCounts(append(counts, raw\_synths))
 6:
            reduced \leftarrow PCADimReduction(augmented)
 7:
            clusters \leftarrow PhenographCluster(reduced)
 8:
            P \leftarrow \text{AssignP-Values}(N, \text{length}(raw\_synths), clusters)
 9:
            for j = 1, \ldots, N do
10:
                doublet[i, j] \leftarrow P[j] \ge 0.99
11:
                                                                     \triangleright Call cell j a doublet this run if \ge 0.99.
            end for
12:
        end for
13:
        labels \leftarrow N length array of zeros
14:
        for j = 1, ..., N do
15:
            if [CountTrue(doublet[:,j])/length(doublet[:,j])] \ge 0.9 then
16:
17:
                labels[j] \leftarrow 1
                                                        \triangleright Cell j was called doublet on at least 90% of runs.
            end if
18:
19:
        end for
        return labels
20:
21: end function
```