# Smashing single cells into k-mer sketches

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# **Abstract**

Single-cell RNA-sequencing is a powerful technology for identifying novel and known cell types, however its power is limited to organisms with well-annotated genomes. We demonstrate the utility of using annotation-agnostic methods which quantify cell-cell similarity using k-mer profiles. We benchmark a few methods and demonstrate the utility of converting cell types from mouse to human and back, and compare to using purely 1:1 mapped orthologous genes.

# Introduction

There are a predicted 8.7 million Eukaryotic species on earth [1], yet only 14% (1,233,500) have been catalogued and 0.000002% (200/8,700,000 = 2.3e-08) have genomes present in ENSEMBL Assemblies (as of ENSEMBL 98 – September 2019 release) [2]. And yet, the genome sequence is not enough. To truly understand the diversity of life on this planet, we need to determine not just the DNA blueprints of life, but understand the instantiation of the DNA, the cell types of the species. While sequencing DNA gives a quantitative measure of the nucleotide differences, it does not inform the functional strategies that change from DNA modifications due to speciation events . As new species can be defined by a new cell type. For example, the existence of a single cell type, the stinging cell called a "Cnidocyte" [3], a single-celled biological weapon, defines the phylum Cnidaria. Thus, entire clades, not only species, can be defined by the introduction of an additional cell type or state.

Organizations of existing cell states can also define novel organismal structures. For example, different physical organizations of similar cell types generate different genitalia in amniotes when comparing mammals to reptiles [4]

Determining common gene ancestry ("orthology") is a difficult problem. Many approaches exist, reviewed by [5,6,7]. Generally, the approaches are structured in this way: (1) find orthologous groups of genes, (2) build gene trees, (3) build species trees, and (4) assign orthologs, as described in a recent approach (Orthofinder) [8]. In this approach, we are not interested in exactly reconstructing the species or gene trees, but rather inferring function based on cell type transcriptomes. Instead of exactly building the gene trees, we subset the protein-coding sequences into peptide words, and reencode to lossy peptide encodings.

Determining common ancestry of cell types ("orthologous cell types") [9,10] is an additional difficult problem. Comparative transcriptomics begins with finding a common feature set for embedding molecular profiles across divergent species into a common space. Many researchers take the approach of using one-to-one orthologous genes [Cite: brawand2011, CCA, LIGER, Scanorama, basically all the single cell "alignment" packages], others use clusters of orthologous groups [11], others map reads onto a common genome derived from whole-genome alignment [cite: recent primate brain paper from Barbara Treutlein], or map onto native genomes [12] and re-annotate using a tool such as Comparative Annotation Toolkit [13].

k-mers have been proposed for comparing single cells [14] as they are a fast, simple way to create cell-cell similarities. However, the work so far has focused on using annotated organisms and not cross-species analyses.

We aim to find "orthologous reads" across species' transcriptomes. By representing each species' transcriptome as the set of k-mers, we can unbiasedly compare transcriptomes in an orthologous space without the need for knowing the orthologous genes ahead of time, or even the need for a reference genome. Additionally, we do not need to reduce the signal to only the genes with a 1:1 orthologous match.

# **Methods**

Methods go here.

# **Experimental**

**Primate brain organoid protocols** 

We did things. One sentence per line. Prefer DOI for references, but for Biorxiv use the URL. DOI example: [5]. Biorxiv example: [8]. Multiple citations per line example: [5,8].

# Single-cell capture of primate brain organoids

We did things. One sentence per line. Prefer DOI for references, but for Biorxiv use the URL. DOI example:  $[\underline{5}]$ . Biorxiv example:  $[\underline{8}]$ . Multiple citations per line example:  $[\underline{5},\underline{8}]$ .

# Long read library prep

We did things. One sentence per line. Prefer DOI for references, but for Biorxiv use the URL. DOI example:  $[\underline{5}]$ . Biorxiv example:  $[\underline{8}]$ . Multiple citations per line example:  $[\underline{5},\underline{8}]$ .

# Short read library prep

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# Sequencing

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# Computational

# k-mer comparison of orthologous genes

We used ENSEMBL version 97. We did things. One sentence per line. Prefer DOI for references, but for Biorxiv use the URL. DOI example: [5]. Biorxiv example: [8]. Multiple citations per line example: [5,8].

### **Extraction of putative coding reads from RNA-seq**

We did things. One sentence per line. Prefer DOI for references, but for Biorxiv use the URL. DOI example: [5]. Biorxiv example: [8]. Multiple citations per line example: [5,8].

# bam2fasta conversion

The .bam file generated by the Drop-seq [15] pipeline for the different primates in this study are in the order of 6-12 GB. The Drop-seq .bam files so obtained can attribute to few limitations as discussed below. Firstly, loading them in memory all at once would require a lot of RAM depending on how the program will allocate memory for different data typed tags in the .bam file. Secondly, if Drop-seq data is not accompanied by a barcodes file to filter the .bam file on, it would mean we would have to recursively go through the alignments in the bam file and deduce alignments with higher quality and combine sequences with already exisiting barcodes. This would need a look up dictionary to be updated as it loops through the alignments in the .bam file and would search the look up dictionary as it updates the barcodes. In conclusion, this is a very memory intensive process that seemed to fail on even machines with 2TB RAM.

Hence we propose a method that could work on a computer with lesser RAM and not cause computer hangups. We released an open source pypi package for the same [16]. The package contains solution for the above discussed problem by sharding the .bam file into chunks of smaller .bam files and stores them in the machine's temporary folder, e.g. /tmp . The chunk size of the .bam file is a

tunable parameter that can be accessed with --line\_count; by default it is 1500 alignment lines. This process is done serially by iterating through the alignments in the .bam file, using pysam, a Python wrapper around samtools [17]. Now we employ a MapReduce [18] approach to the temporary .bam files to obtain all the reads per cell barcode in a .fasta file. In the "Map" step, we distribute the computation i.e parsing the barcode, determining the quality of the read, and if alignment is not duplicated, in parallel across multiple processes on the temporary shards of .bam files. These bam shards create temporary .fasta files that contain for each read: the cell barcode, unique molecular identifier (UMI), and the aligned sequence. There might be a cell barcode that would be present in different chunks of these sharded .bam files. As a result we would have multiple temporary .fasta files for the same barcodes. We implemented a method to find the unique barcodes based on these temporary . fasta file names and then assigning each of the unique barcodes all the temporary barcode .fasta files created by different .bam shards in a dictionary. In the "Reduce" step, we concatenate of strings of temporary .fasta file names, hence its memory consumption is less than it would be if appending to a list. These temporary .fasta files are then combined to one .fasta file per barcode by concatenating all the sequences obtained from different .fasta files. The concatenation of all sequences for each of the unique barcodes is also then parallelized to use multiple processes. For each of the cell barcodes, there is an option to obtain valid cell barcodes, based on the UMI count per cell barcode. For our datasets we have set the minimum number of UMIs per cell barcode to 1000, a common threshold. The minimum number of UMIs per cell barcode can be customized with the flag --min-umi-per-barcode. The computational resources and time taken for processing is as shown in Table [1].

**Table 1:** Human primate species bam file here is from a brain organoid for human

Primate	BAM file size(GB)	Time(hrs)	RAM(GB)	Processes
Human	12	7	16	32
Orangutan	9	4	16	32
Chimp	9	4	16	32

This method primarily gives us time performance improvement. It reduces time from days or just process running out of memory to hours. Depending on the size of .bam file and resources of the cluster/computer it can be further reduced.

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### Results

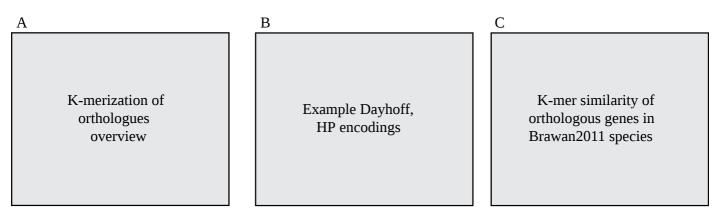
To determine whether short segments of sequences could detect gene orthologues, we k-merized orthologous genes derived from the ENSEMBL version 97 [19] COMPARA database [20] (Figure [1]). We compared human protein sequences to orthologous chimpanzee, mouse, (orangutan, bonobo, gorilla, macaque, opossum, platypus, chicken) protein sequences, as these are species used in [21]. As a background, we randomly chose 10 non-orthologous genes relative to the human gene. In addition to k-merizing the protein-coding sequence, we also re-encoded the protein-coding sequence into Dayhoff [22] and hydrophobic-polar encodings [23], show in Table [2].

**Table 2:** Dayhoff and hydrophobic-polar encodings are a reduced amino acid alphabet allowing for permissive cross-species sequence comparisons. For example, the amino acid sequence SASHAFIERCE would be Dayhoff-encoded to bbbdbfecdac, and HP-encoded to phpphhhpppp.

Amino acid Property Dayhoff Hydrophobic-polar (HP)
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Amino acid	Property	Dayhoff	Hydrophobic-polar (HP)
С	Sulfur polymerization	a	р
A, G, P, S, T	Small	b	A, G, P: h
			S,T: p
D, E, N, Q	Acid and amide	С	р
H, K, R	Basic	d	р
I, L, M, V	Hydrophobic	е	h
F, W, Y	Aromatic	f	h

k-mer size may be tuned to find an "optimal" length of protein domains across the tree of life. Protein domain lengths follow a power law distribution where proteins with more domains, have shorter domains, whereas proteins with fewer domains, have fewer but longer domains [24,25].



**Figure 1:** Figure 1.

We found that, consistent with previous knowledge, that 1:1 orthologues had higher k-mer similarities as determined by the Jaccard Index.

Additionally, more recently diverged genes had higher k-mer similarity as well.

Extract coding sequence using human proteins, k-merize coding sequence and minhash, com pare samples/cells across minhashes

В

kNN graph of Brawand2011 MinHashes kNN graph of Brawand2011 Gene expression

K-mers driving similarity in brawand2011

Are the k-mers from unmapped reads or unannotated genes?

Figure 2: Figure 2.

### **Outlines**

Figure 1 outline

- Kmers can approximate orthologies
  - Jaccard similarity of orthologues is higher than non-orthologues
  - Benchmarking using https://orthology.benchmarkservice.org/cgi-bin/gateway.pl
  - Finding orthologues
    - Gold standard
      - ENSEMBL COMPARA
      - Quest for Orthologs consortium, Altenhoff, A. M., Boeckmann, B., Capella-Gutierrez, S., Dalquen, D. A., DeLuca, T., et al. (2016). Standardized benchmarking in the quest for orthologs. Nature Methods, 13(5), 425–430. http://doi.org/10.1038/nmeth.3830 [5]
    - Orthologous groups/Conserved Domain Database [26]
- Kmers can find correct reading from of RNA-seq reads
  - Human peptides → human RNAseq
  - Human peptides → chimp RNAseq
  - Human peptides → mouse RNAseq
- Kmers can find only transcription factor reads of TFs from RNA-seq reads
  - Human TFs → human RNAseq
  - Human TFs → chimp RNAseq
  - Human TFs → mouse RNAseq
- Overview of kmermaid pipeline
  - Comparison of tissue across species
    - Partition reads to coding/noncoding bins
    - MinHash the Dayhoff-encoded coding sequences
    - Jaccard similarity on the MinHashes

# Figure 2 outline

References for Primate Brain development

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### No need for 1:1 orthology

Gene expression evolution through duplications

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Don't need to normalize gene expression counts since we just have presence/absence

# Full transcript analyses

• Brain Organoid - Droplet + PacBio data

# Correlated evolution of celltypes?

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- · Mammalian decidual cell

# Cell type evolution

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# Figure 3 - long evolutionary distances with HP encoding?

# Metazoan body plan formation

- Early development in Cnidarians/Hydra [37,38]
- sponges and others [38]
- planaria [39,40],
- drosophila [41]
- zebrafish [42,43,44],
- mouse [45]

# Figure 4 - What features are k-mers able to pick up that mapping doesn't?

- Which reads are found to have coding features but didn't map to the genome?
  Do these features map to novel genes or gene fusions?

# **Discussion**

Conclusions and future directions go here.

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