

Tim

### **A strategy for generating negative control pairs: version # 2**

I describe another, more workable (and simpler) strategy for generating negative control pairs. Again, define the following variables.

- The number of negative control gRNAs  $N_{\text{grna}}$ . Label the NT gRNAs  $1, 2, \dots, N_{\text{grna}}$ .
- The  $N_{\text{cell}} \times N_{\text{gene}}$  matrix of gene expressions and the  $N_{\text{cell}}$ -dimensional gRNA-to-cell assignment vector.
- The number of pairs to generate  $N_{\text{pairs}}$ .
- The undercover group size  $k \leq N_{\text{grna}}/2$ .
- The minimum number of treatment cells  $N_{\text{trt}}$  and control cells  $N_{\text{ctrl}}$  needed for a pair to pass pairwise QC.

We proceed in several steps.

**Step 1: Tabulate the number of cells with nonzero expression for each individual NT gRNA and gene.** First, we compute an  $N_{\text{grna}} \times N_{\text{gene}}$  matrix  $M$ , where entry  $(i, j)$  is the number of cells containing NT gRNA  $i$  with nonzero expression of gene  $j$ . We easily can construct this matrix either in memory or out-of-core by summing over columns of the gene expression matrix.

**Step 2: Determine if it is feasible to enumerate the possible undercover gRNA groups.** We check the value of  $N_{\text{possible-groups}} := \binom{N_{\text{gene}}}{k}$ . If  $N_{\text{possible-groups}}$  is a huge number (e.g.,  $\binom{100}{50} \approx 10^{30}$ ), then it is not possible to enumerate the possible undercover gRNA groups. If  $N_{\text{possible-groups}}$  is small, by contrast (e.g.,  $\binom{9}{2} = 36$ ), then it is possible to enumerate the possible undercover gRNA groups. We check if  $N_{\text{possible-groups}}$  exceeds some pre-defined threshold (e.g., 20,000), carrying out a different routine in either case. If  $N_{\text{possible-groups}} \leq 20,000$ , then we proceed to step 3a. Otherwise, we proceed to step 3b.

**Step 3a: Enumerate the possible undercover gRNA groups.** If

$N_{\text{possible-groups}} \leq 20,000$ , we enumerate the possible undercover gRNA groups. We map each possible undercover gRNA group to a length- $k$  vector of integers sorted in increasing order, where the integers represent individual NT gRNAs. For example, the undercover gRNA group containing NT gRNAs 2, 3, and 7 (arbitrarily labeled) would be mapped to the vector [ 2, 3, 7 ]. We then generate the entire set of  $N_{\text{possible-groups}}$  length- $k$  vectors containing integers in the range  $\{1, \dots, N_{\text{grna}}\}$ . We store these vectors in an ordered list  $\mathbf{x}$ . We also set  $N_{\text{grna-groups}} = N_{\text{possible-groups}}$ .

**Step 3b: Sample a set of possible undercover gRNA groups.** If  $N_{\text{possible-groups}} > 20,000$ , then we do not attempt to enumerate the entire set of possible undercover gRNA groups. Instead, we sample a set of undercover gRNA groups. We proceed as follows. First, we estimate the fraction of undercover gRNA-gene pairs (of group size  $k$ ) that passes QC. We do this by pairing a randomly generated undercover gRNA group to a randomly selected gene and checking if that pair passes the pairwise QC threshold. We sample (with replacement) a large number (e.g., 5,000) undercover gRNA-gene pairs in this way, producing an estimate  $\hat{p}$  of the fraction of undercover gRNA-gene pairs that passes QC. We then set the number of gRNA groups to sample  $N_{\text{grna-groups}}$  to

$$N_{\text{grna-groups}} = \frac{c \cdot N_{\text{pairs}}}{\hat{p} N_{\text{genes}}},$$

where  $c > 1$  is a number that ensures we sample a *conservative* number of gRNA groups (i.e., more than we need). Finally, we sample  $N_{\text{grna-groups}}$  gRNA groups by sampling from the set of length  $k$  vectors containing integers in the range  $\{1, \dots, N_{\text{grna}}\}$  via membership checking sampling.\* We store these  $N_{\text{grna-groups}}$  vectors in an ordered list  $\mathbf{x}$ .

\* To be more specific, we sample  $N_{\text{grna-groups}}$  gRNA groups as follows. We initialize an empty set (implemented as a hash table)  $\mathcal{D}$ . We then construct a length- $k$  sample from the set  $\{1, \dots, N_{\text{grna}}\}$  via Fisher-Yates sampling and sort the resulting vector. Finally, we check for inclusion of this vector in  $\mathcal{D}$ . If the vector already is in  $\mathcal{D}$ , we proceed to the next iteration. Otherwise, we add this vector to  $\mathcal{D}$ . We conclude this process when the number of elements in  $\mathcal{D}$  is equal to  $N_{\text{grna-groups}}$ . In the rare case that  $N_{\text{grna-groups}} \geq N_{\text{possible-groups}}$ , we can construct the gRNA groups via enumerating over all combinations, as in step 3a.

**Step 4: Sample without replacement from the set of undercover gRNA group-gene pairs.** The final step is to sample a set of undercover gRNA group-gene pairs without replacement. Recall that step 3 yields a list  $\mathbf{x}$  of undercover gRNA groups of length  $N_{\text{grna-groups}}$ . (This is true whether we have carried out step 3a or step 3b). There are thus  $N_{\text{grna-group}} \cdot N_{\text{gene}}$  pairs that we could sample. We map each gRNA group-gene pair in this set of pairs to an integer in the set  $\{1, \dots, N_{\text{gene}} \cdot N_{\text{grna-group}}\}$ . The map is defined as follows: for an integer  $i \in \{1, \dots, N_{\text{gene}} \cdot N_{\text{grna-group}}\}$ , we carry out the integer division

$$\text{grna\_group\_idx} = i \% N_{\text{gene}},$$

which defines a gRNA group index. Next, we compute the remainder of this division

$$\text{gene\_idx} = i \% \% N_{\text{gene}}$$

to compute a gene index. Through this map, sampling without replacement from the set of integers  $\{1, \dots, N_{\text{gene}} \cdot N_{\text{grna-group}}\}$  is identical to sampling without replacement from the set of undercover gRNA group-gene pairs.

If the number of pairs to sample  $N_{\text{pairs}}$  exceeds the number of pairs that we possibly could sample  $N_{\text{gene}} \cdot N_{\text{grna-group}}$ , then we iterate through the pairs one-by-one, checking if the pair passes pairwise-QC and, if so, adding it to the set pairs to return. If, on the other hand, the number of pairs to sample  $N_{\text{pairs}}$  is less than  $N_{\text{gene}} \cdot N_{\text{grna-group}}$ , we sample without replacement from the set  $\{1, \dots, N_{\text{gene}} \cdot N_{\text{grna-group}}\}$ , discarding those pairs that do not pass QC. (We can implement this final sampling without replacement step via Fisher-Yates sampling or sparse Fisher-Yates sampling.)

**Background on without replacement sampling** See the preprint “Simple, Optimal Algorithms for Random Sampling without Replacement” (Ting, 2021) for descriptions of Fisher-Yates, sparse Fisher-Yates, and membership checking algorithms for without replacement sampling.