

Figure 1 (scRNA & 20 Newsgroups): Wall-clock runtime versus scaling with k and n for the SCRNA (L1 distance) and 20 Newsgroups (cosine distance) datasets. We observe speedups of up to  $6.4 \times$  and  $8.5 \times$  speedup when examining the scaling of each algorithm with k and  $3.2 \times$  and  $7 \times$  speedup for scaling of n for the two datasets, respectively.

Table 1: Average Runtime Speedup Summary							
Dataset	BUILD + SWAP	SWAP only	BUILD only				
MNIST CIFAR SCRNA 20 Newsgroups	$\begin{array}{c} \times 4.75 \mid \times 8.23 \mid \times 10.77 \\ \times 5.05 \mid \times 9.24 \mid \times 12.52 \\ \times 3.95 \mid \times 5.12 \mid \times 6.81 \\ \times 5.19 \mid \times 7.62 \mid \times 9.18 \end{array}$	$\begin{array}{c} \times 6.20 \mid \times 9.33 \mid \times 12.18 \\ \times 6.29 \mid \times 11.8 \mid \times 15.03 \\ \times 4.74 \mid \times 6.01 \mid \times 7.94 \\ \times 6.67 \mid \times 8.98 \mid \times 10.27 \end{array}$	$\begin{array}{c} \times 1.58 \mid \times 1.83 \mid \times 2.16 \\ \times 2.49 \mid \times 3.27 \mid \times 3.65 \\ \times 2.67 \mid \times 2.91 \mid \times 3.44 \\ \times 2.52 \mid \times 2.90 \mid \times 3.16 \end{array}$				

**Table 1** (Average Speedup Table): Wall-clock speedup of BanditPAM++ compared to BanditPAM on the four datasets MNIST, CIFAR, SCRNA, and 20 Newsgroups. Results were averaged over four different dataset sizes n = 10k, 20k, 30k, 50k for settings BUILD + SWAP, SWAP only, and BUILD only. The BUILD only setting leverages permutation-invariant caching only, whereas the other two settings also leverage Virtual Arms. The three speedup values in each cell correspond to experiments where k = 5, 10, and 15 respectively.

Table 2:	Relative Los	s with	Varying $\delta$
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Dataset	$10^{-2}$	$10^{-3}$	$10^{-5}$	$10^{-10}$
MNIST	1.0	1.0	1.0	1.0
20 Newsgroups	1.0	1.0	1.0	1.0

**Table 2** (Loss with  $\delta$ ): Loss of BanditPAM++ over BanditPAM with  $\delta$  values ranging from  $10^{-2}$ ,  $10^{-3}$ ,  $10^{-5}$ , and  $10^{-10}$ . BanditPAM++ has the exact same clustering loss with BanditPAM for various values  $\delta$ .



Figure 2 (Clustering Loss with Varying T): Clustering loss with increasing T for the MNIST and CIFAR datasets for k = 5 and k = 10. Beyond T = k, the loss shows very little change. BanditPAM++ and BanditPAM have the same loss for all T making them trace the exact same optimization trajectory with increasing T.