A Appendix: Proof of Theorem 1

We first show that the estimate is unbiased. Indeed, for every $i \neq j$ we can rewrite L(z) as $\mathbb{E}_{\pi} \ell_{\pi(i),\pi(j)}(z)$. Therefore,

$$L(z) = \frac{1}{k^2 - k} \sum_{i \neq j \in [k]} L(z) = \frac{1}{k^2 - k} \sum_{i \neq j \in [k]} \mathbb{E}_{\pi} \ell_{\pi(i), \pi(j)}(z) = \mathbb{E}_{\pi} L_{\pi}(z) ,$$

which proves that the multibatch estimate is unbiased.

Next, we turn to analyze the variance of the multibatch estimate. let $I \subset [k]^4$ be all the indices i, j, s, t s.t. $i \neq j, s \neq t$, and we partition I to $I_1 \cup I_2 \cup I_3$, where I_1 is the set where i = s and j = t, I_2 is when all indices are different, and I_3 is when i = s and $j \neq t$ or $i \neq s$ and j = t. Then:

$$\begin{split} & \underset{\pi}{\mathbb{E}} \|\nabla L_{\pi}(z) - \nabla L(z)\|^{2} = \frac{1}{(k^{2} - k)^{2}} \underset{\pi}{\mathbb{E}} \sum_{(i, j, s, t) \in I} (\nabla \ell_{\pi(i), \pi(j)}(z) - \nabla L(z)) \cdot (\nabla \ell_{\pi(s), \pi(t)}(z) - \nabla L(z)) \\ & = \sum_{r=1}^{d} \frac{1}{(k^{2} - k)^{2}} \sum_{q=1}^{3} \sum_{(i, j, s, t) \in I_{q}} \underset{\pi}{\mathbb{E}} (\nabla_{r} \ell_{\pi(i), \pi(j)}(z) - \nabla_{r} L(z)) (\nabla_{r} \ell_{\pi(s), \pi(t)}(z) - \nabla_{r} L(z)) \end{split}$$

For every r, denote by $A^{(r)}$ the matrix with $A_{i,j}^{(r)} = \nabla_r \ell_{i,j}(z) - \nabla_r L(z)$. Observe that for every r, $\mathbb{E}_{i \neq j} A_{i,j}^{(r)} = 0$, and that

$$\sum_{r} \mathop{\mathbb{E}}_{i \neq j} (A_{i,j}^{(r)})^2 = \mathop{\mathbb{E}}_{i \neq j} \|\nabla \ell_{i,j}(z) - \nabla L(z)\|^2.$$

Therefore,

$$\mathbb{E}_{\pi} \|\nabla L_{\pi}(z) - \nabla L(z)\|^{2} = \sum_{r=1}^{d} \frac{1}{(k^{2} - k)^{2}} \sum_{q=1}^{3} \sum_{(i,j,s,t) \in I_{q}} \mathbb{E}_{\pi} A_{\pi(i),\pi(j)}^{(r)} A_{\pi(s),\pi(t)}^{(r)}$$

Let us momentarily fix r and omit the superscript from $A^{(r)}$. We consider the value of $\mathbb{E}_{\pi} A_{\pi(i),\pi(j)} A_{\pi(s),\pi(t)}$ according to the value of q.

- For q = 1: we obtain $\mathbb{E}_{\pi} A^2_{\pi(i),\pi(j)}$ which is the variance of the random variable $\nabla_r \ell_{i,j}(z) \nabla_r L(z)$.
- For q = 2: When we fix i, j, s, t which are all different, and take expectation over π , then all products of off-diagonal elements of A appear the same number of times in $\mathbb{E}_{\pi} A_{\pi(i),\pi(j)} A_{\pi(s),\pi(t)}$. Therefore, this quantity is proportional to $\sum_{p \neq r} v_p v_r$, where v is the vector of all non-diagonal entries of A. Since $\sum_p v_p = 0$, we obtain (using Lemma 1) that $\sum_{p \neq r} v_p v_r \leq 0$, which means that the entire sum for this case is non-positive.
- For q = 3: Let us consider the case when i = s and j ≠ t, and the derivation for the case when i ≠ s and j = t is analogous. The expression we obtain is E_π A_{π(i),π(j)}A_{π(i),π(t)}. This is like first sampling a row and then sampling, without replacement, two indices from the row (while not allowing to take the diagonal element). So, we can rewrite the expression as:

$$\mathbb{E}_{\pi} A_{\pi(i),\pi(j)} A_{\pi(s),\pi(t)} = \mathbb{E}_{i \sim [m] \ j,t \in [m] \setminus \{i\}: j \neq t} \mathbb{E}_{A_{i,j} A_{i,t}}$$

$$\leq \mathbb{E}_{i \sim [m]} \left(\mathbb{E}_{j \neq i} A_{i,j} \right)^2 = \mathbb{E}_{i \sim [m]} (\bar{A}_i)^2,$$
(5)

where we denote $\bar{A}_i = \mathbb{E}_{j \neq i} A_{i,j}$ and in the inequality we used again Lemma 1.

Finally, the bound on the variance follows by observing that the number of summands in I_1 is $k^2 - k$ and the number of summands in I_3 is $O(k^3)$. This concludes our proof.

Lemma 1 Let $v \in \mathbb{R}^n$ be any vector. Then,

$$\mathop{\mathbb{E}}_{s \neq t} [v_s v_t] \le (\mathop{\mathbb{E}}_{i} [v_i])^2$$

In particular, if $\mathbb{E}_i[v_i] = 0$ then $\sum_{s \neq t} v_s v_t \leq 0$.

Proof For simplicity, we use $\mathbb{E}[v]$ for $\mathbb{E}_i[v_i]$ and $\mathbb{E}[v^2]$ for $\mathbb{E}_i[v_i^2]$. Then:

$$\begin{split} \mathbb{E}_{s \neq t} v_s v_t &= \frac{1}{n^2 - n} \sum_{s=1}^n \sum_{t=1}^n v_s v_t - \frac{1}{n^2 - n} \sum_{s=1}^n v_s^2 \\ &= \frac{1}{n^2 - n} \sum_{s=1}^n v_s \sum_{t=1}^n v_t - \frac{1}{n^2 - n} \sum_{s=1}^n v_s^2 \\ &= \frac{n^2}{n^2 - n} \mathbb{E}[v]^2 - \frac{n}{n^2 - n} \mathbb{E}[v^2] \\ &= \frac{n}{n^2 - n} (\mathbb{E}[v]^2 - \mathbb{E}[v^2]) + \frac{n^2 - n}{n^2 - n} \mathbb{E}[v]^2 \\ &\leq 0 + \mathbb{E}[v]^2 \end{split}$$