

Rebuttal Response

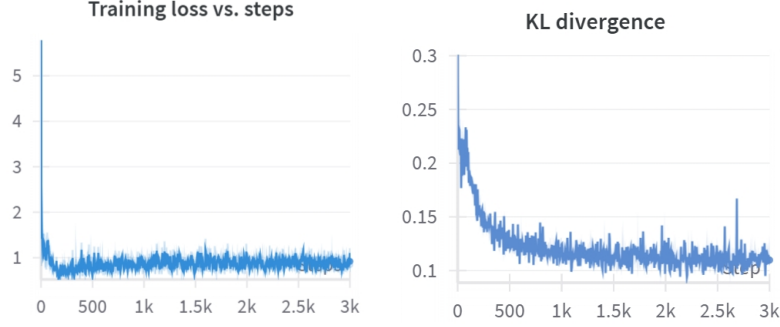


Figure 1. The evolution of the lower-level training loss and KL divergence in training of 10B tokens. Proxy model size: 31M, target LLM size: 410M.

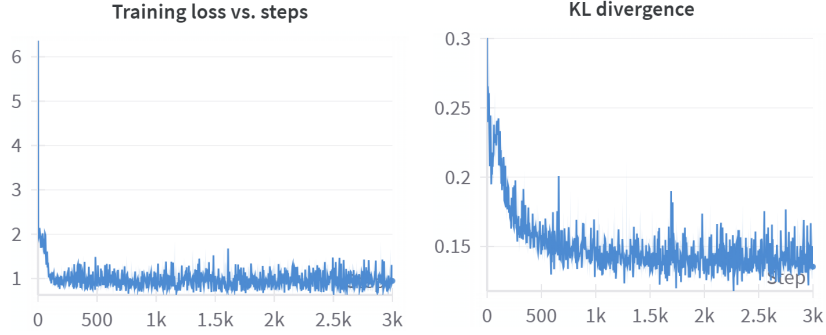


Figure 2. The evolution of the lower-level training loss and KL divergence in training of 10B tokens. Proxy model size: 160M, target LLM size: 410M.

Table 1. Total FLOPs for pretraining 410M/1B target model with 25B tokens.

| Process | #FLOPs $\times 10^{19}$ | Ratio |
|--------------------------------------|-------------------------|----------------|
| BLISS: 410M model, 25B tokens | | |
| Model pretraining | 6.35 | 79.28% |
| Warm up the proxy/score model | 0.07 | 0.87% |
| Bilevel optimization | 0.13 | 1.62% |
| Data influence model inference | 1.53 | 19.10% |
| Total | 8.08 | 100.00% |
| BLISS: 1B model, 25B tokens | | |
| Model pretraining | 17.67 | 90.48% |
| Warm up the proxy/score model | 0.07 | 0.36% |
| Bilevel optimization | 0.261 | 1.34% |
| Data influence model inference | 1.53 | 7.83% |
| Total | 19.53 | 100.00% |

Table 2. Comparison of BLISS with different size of proxy/score model and on zero-shot evaluation over multiple downstream datasets (410M model, 10B tokens) with 20k-step training.

| Method | SciQ | ARC-E | ARC-C | LogiQA | OBQA | BoolQ | HellaSwag | PIQA | WinoGrande | Average |
|------------------------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| BLISS (Pythia-31M) | 65.5 _(1.5) | 40.8 _(1.0) | 23.4 _(1.2) | 27.2 _(1.7) | 29.8 _(2.0) | 58.9 _(0.9) | 36.0 _(0.5) | 67.6 _(1.1) | 53.4 _(1.4) | 44.7 _(1.3) |
| BLISS (Pythia-160M) | 63.8 _(1.5) | 40.8 _(1.0) | 23.4 _(1.2) | 27.5 _(1.8) | 29.8 _(2.0) | 51.3 _(0.9) | 38.3 _(0.5) | 67.6 _(1.1) | 50.4 _(1.4) | 44.1 _(1.3) |
| BLISS (Pythia-31M without sigmoid) | 62.6 _(1.5) | 41.0 _(1.0) | 24.0 _(1.2) | 26.4 _(1.7) | 30.4 _(2.1) | 53.4 _(0.9) | 39.5 _(0.5) | 68.3 _(1.1) | 52.2 _(1.4) | 44.2 _(1.3) |

Table 3. Comparison of methods on zero-shot evaluation over multiple downstream datasets (410M model, 15B tokens). BLISS-org denotes the original algorithm, and BLISS[†] is a variant which uses different initialization method for the score model.

| Methods (#FLOPs $\times 10^{19}$) | SciQ | ARC-E | ARC-C | LogiQA | OBQA | BoolQ | HellaSwag | PIQA | WinoGrande | Average |
|------------------------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| BLISS-org | 67.7 _(1.5) | 41.7 _(1.0) | 23.6 _(1.2) | 25.8 _(1.7) | 28.4 _(2.0) | 56.0 _(0.8) | 39.7 _(0.5) | 68.7 _(1.1) | 53.2 _(1.4) | 44.9 _(1.3) |
| BLISS [†] | 65.2 _(1.5) | 41.6 _(1.0) | 23.4 _(1.2) | 27.1 _(1.7) | 29.8 _(2.0) | 57.5 _(0.8) | 34.9 _(0.5) | 67.7 _(1.1) | 53.5 _(1.4) | 44.5 _(1.3) |