

# Programming every example: Lifting pre-training data quality like experts at scale

Fan Zhou<sup>1,3\*</sup>, Zengzhi Wang<sup>1,3\*</sup>, Qian Liu<sup>2</sup>, Junlong Li<sup>1</sup>, Pengfei Liu<sup>1,3</sup>

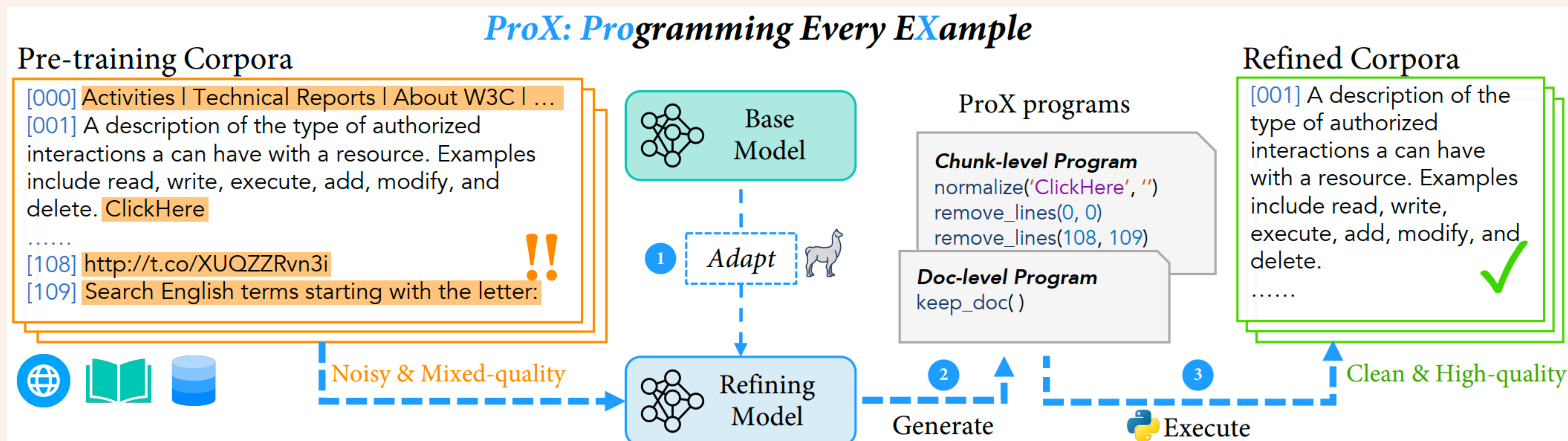
<sup>1</sup>Shanghai Jiao Tong University <sup>2</sup>Sea AI Lab <sup>3</sup>GAIR Lab



**TL;DR** Still worried about the potential noise and the low quality of your rule-cleaned pre-training corpora? **Try ProX!** In ProX, we use 0.3B LLMs to seamlessly refine your pre-training dataset, providing a clean start for your LLM training. ProX serves as a LLM-driven framework which generate **sample-wise, executable** cleaning programs to clean **each and every one** of your pre-training data samples.

- Our 1.7B model, trained on **ProX** corpus with **50B** tokens training, performs on par with **TinyLlama-1.1B** which is trained on **3T** tokens.
- Continual pre-training of **CodeLlama-7B** on **OpenWebMath** with **10B** tokens refined by ProX matches **Llemma-7B**, also pre-trained from **CodeLlama-7B**, trained on **200B** tokens.
- We release “ProX” series of pre-training dataset including **>600B** general pre-training data, and **5B** math corpus.

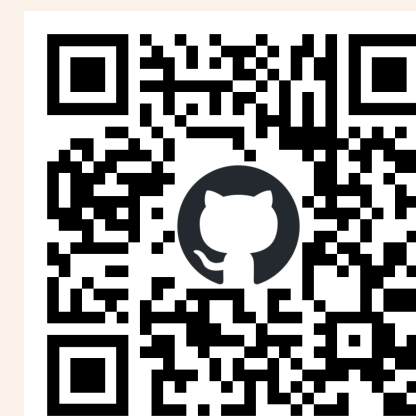
## How ProX Works? *Program Design* *LLM Annotation* *SFT on Tiny LLMs*



Due to visa restrictions, the authors are not able to attend in person. If you have any questions, please also feel free to reach out via the QR codes!



Github



Huggingface

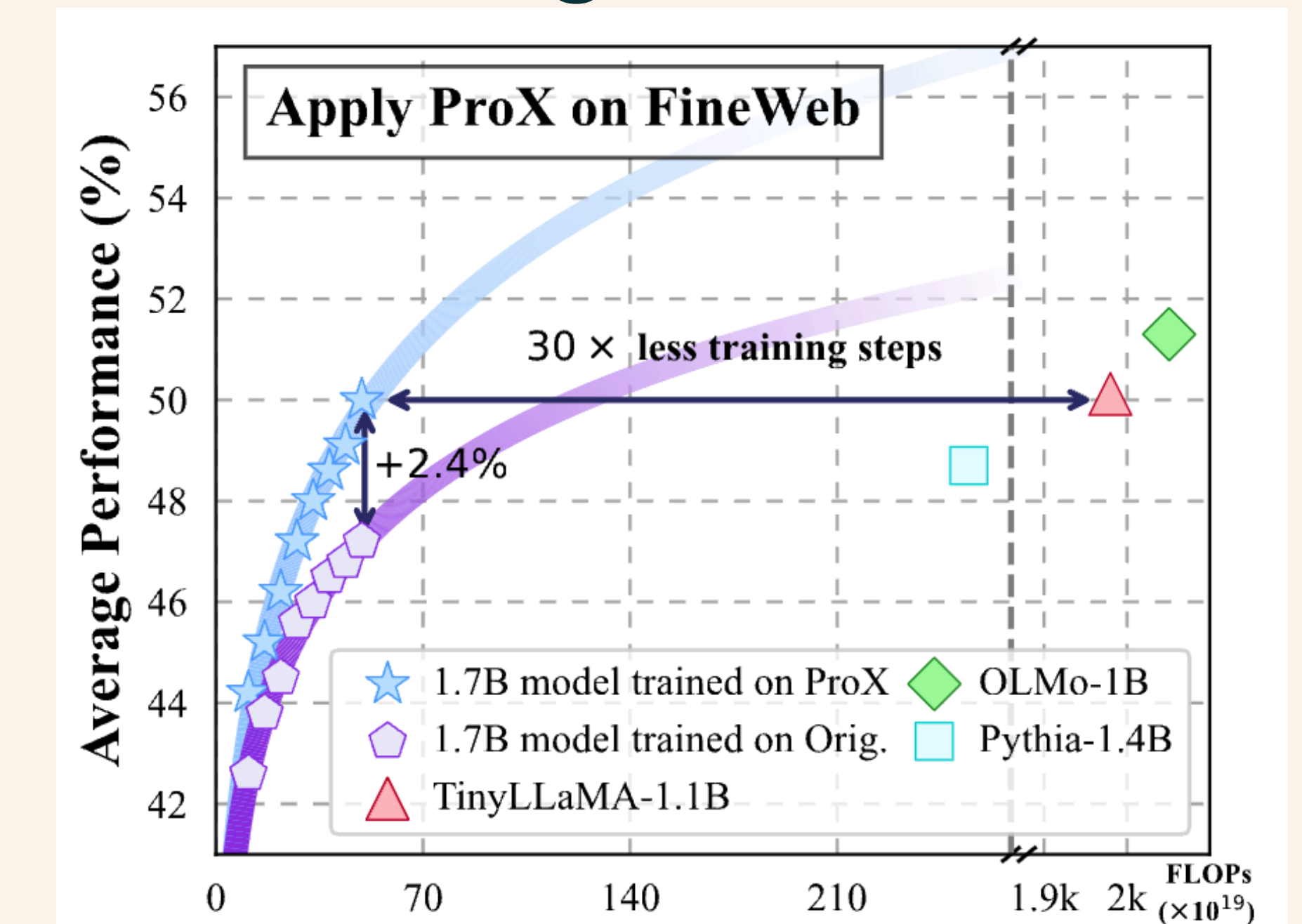


X.COM

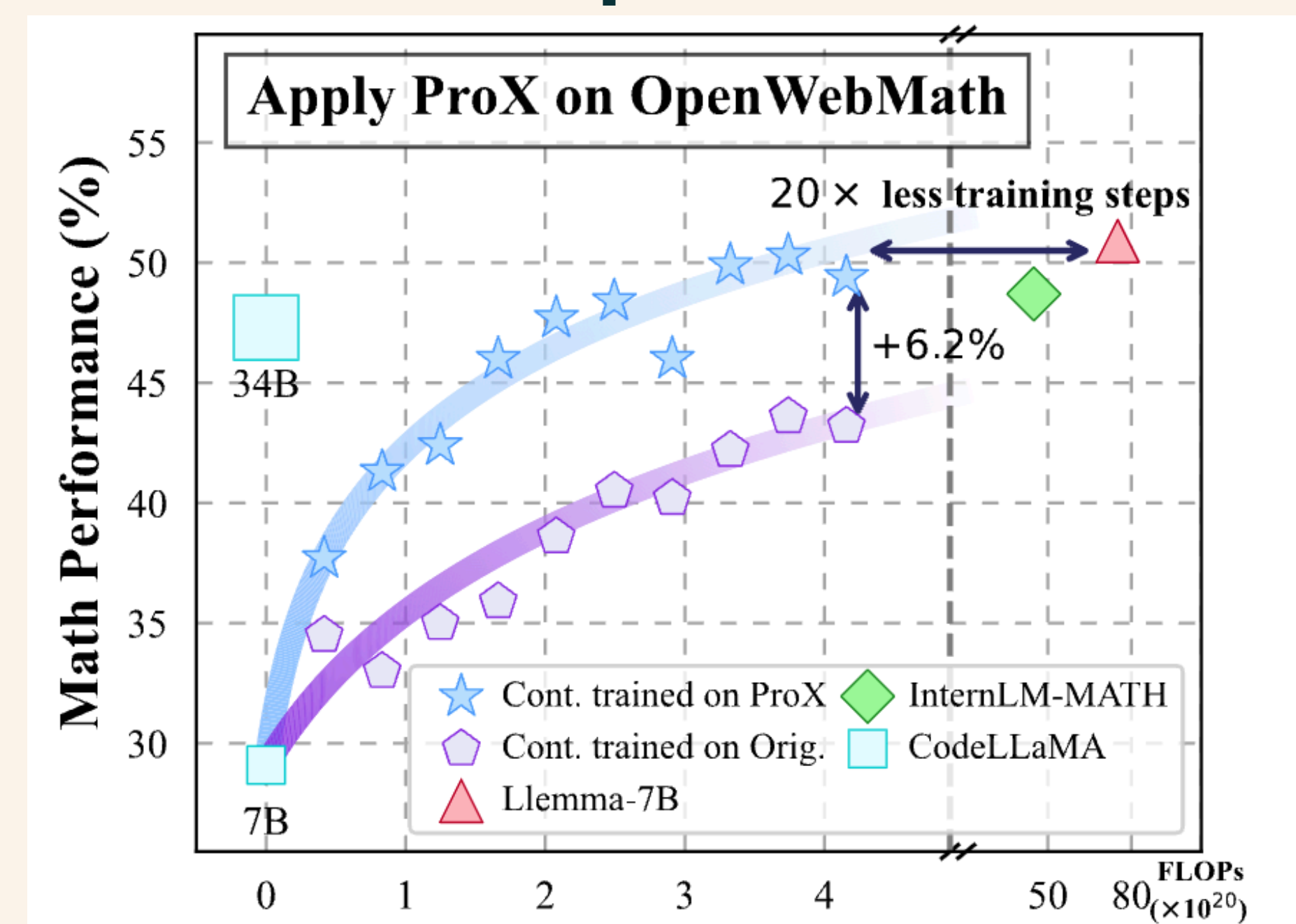


## Experimental Results

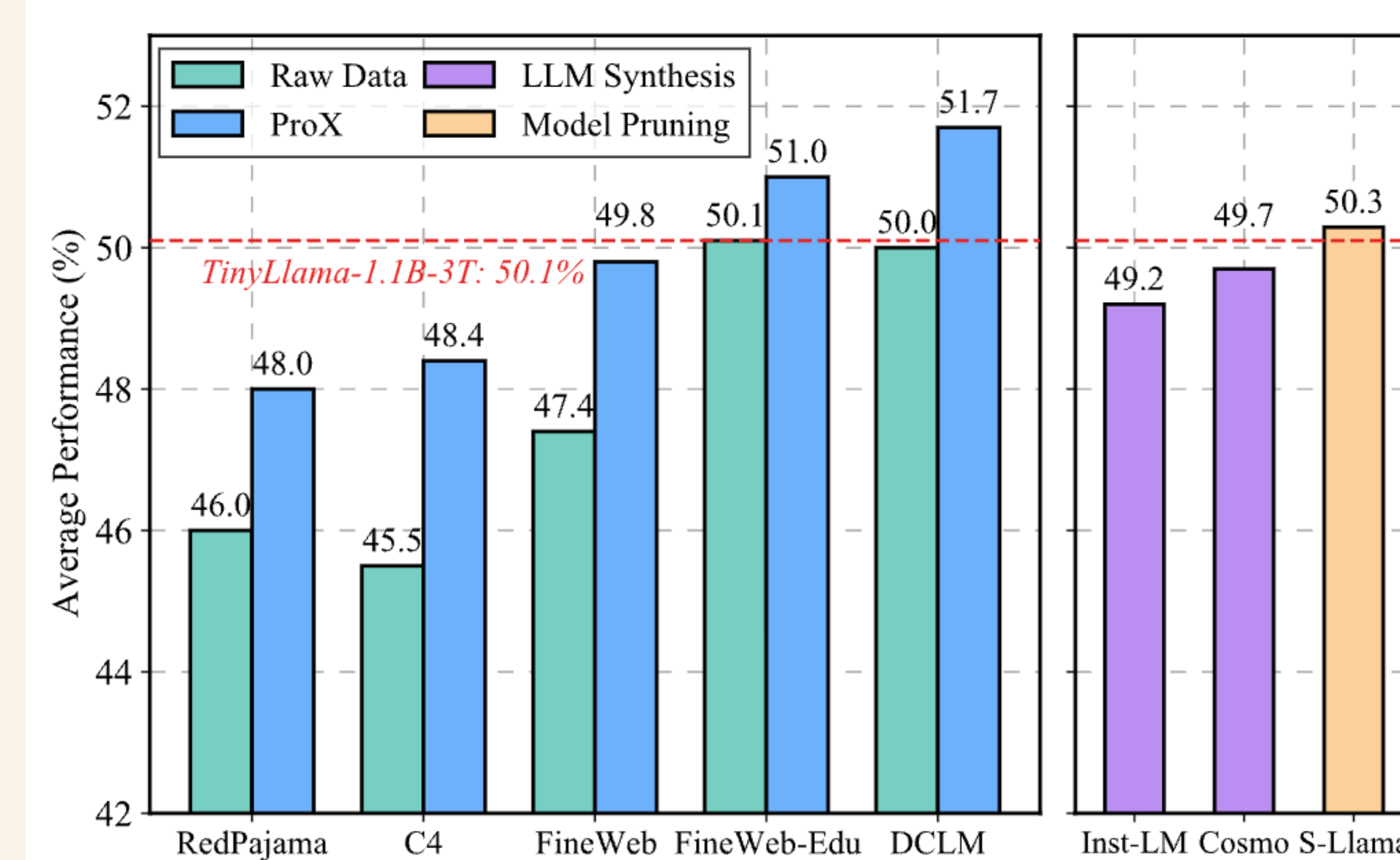
It works on general domain.



It works on specific domain.



And also better than .....



With less overall compute!

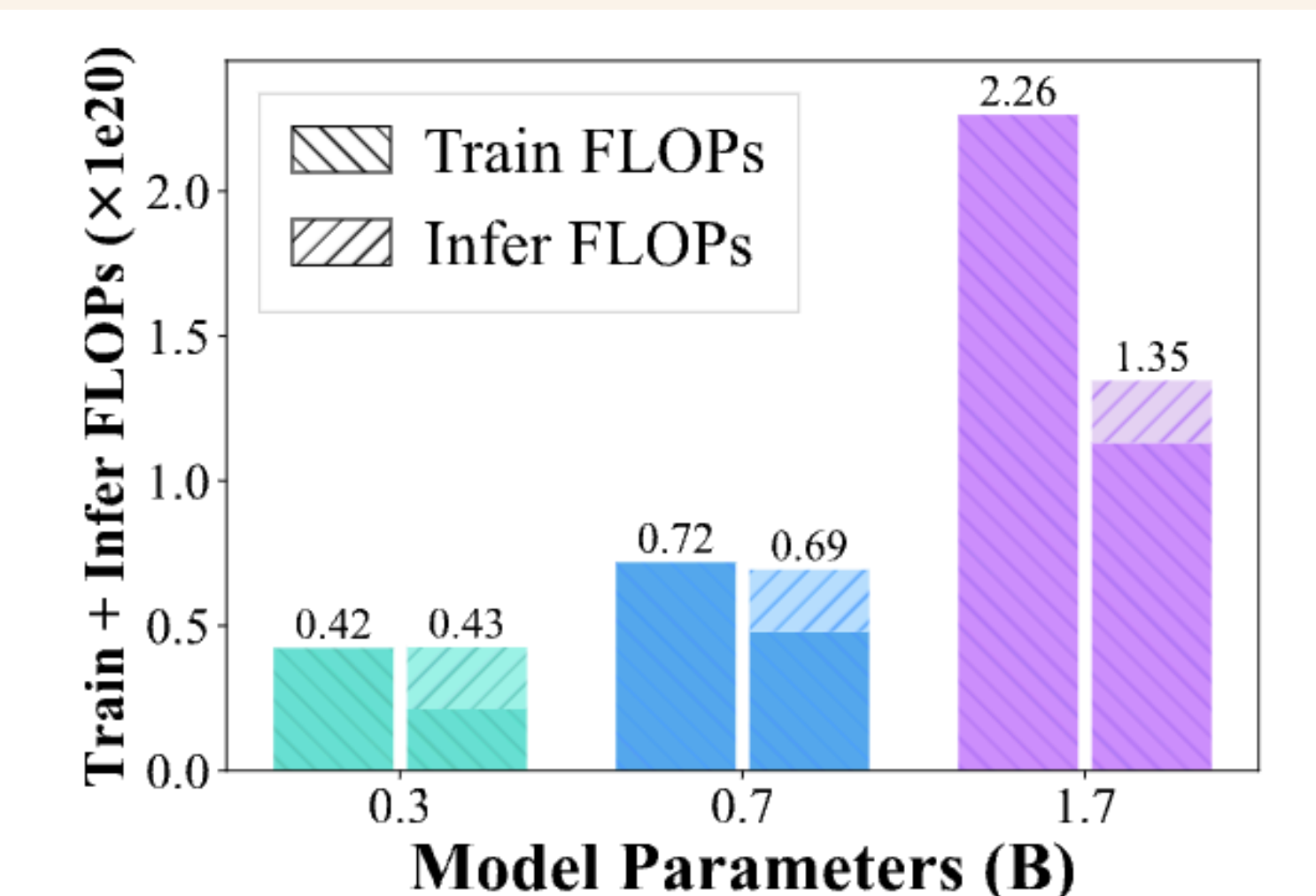


Figure 8: FLOPs comparison for comparable downstream performance with/without PROX refining: 0.3B(Avg.Perf = 40.5), 0.7B (41.6), and 1.7B (42.9).<sup>2</sup>