FUNDAMENTALS OF DEEP LEARNING FOR MULTI-GPUS LAB 3, PART 1: SCALING THE BATCH SIZE



CAN WE INCREASE THE BATCH SIZE INDEFINITELY?

IN TERMS OF IMAGES / SECOND?



Kurth, T., Treichler, S., Romero, J., Mudigonda, M., Luehr, N., Phillips, E., ... & Houston, M. (2018, November). Exascale deep learning for climate analytics. In Proceedings of the International Conference for High Performance Computing, Networking, Storage, and Analysis (p. 51). IEEE Press. <u>arXiv:1810.01993</u>



IN TERMS OF STEPS TO CONVERGENCE? There are limits



Shallue, C. J., Lee, J., Antognini, J., Sohl-Dickstein, J., Frostig, R., & Dahl, G. E. (2018). Measuring the effects of data parallelism on neural network training. arXiv:1811.03600

IN TERMS OF STEPS TO CONVERGENCE?

There are limits







LARGE MINIBATCH AND ITS IMPACT ON ACCURACY

Naïve approaches lead to degraded accuracy



You, Y., Zhang, Z., Hsieh, C., Demmel, J., & Keutzer, K. (2017). ImageNet training in minutes. arXiv: 1709.05011



Naïve approaches lead to degraded accuracy



Hoffer, E., Hubara, I., & Soudry, D. (2017). Train longer, generalize better: closing the generalization gap in large batch training of neural networks. <u>arXiv:1705.08741</u>



Why? Generalization and flatness of minima?



Keskar, N. S., et al. (2016). On large-batch training for deep learning: Generalization gap and sharp minima. arXiv:1609.04836

Why does it happen? Noise in the gradient update.



Keskar, N. S., et al. (2016). On large-batch training for deep learning: Generalization gap and sharp minima. <u>arXiv:1609.04836</u>



Figure 3: The 1D and 2D visualization of solutions obtained using SGD with different weight decay and batch size. The title of each subfigure contains the weight decay, batch size, and test error.

Li, H., Xu, Z., Taylor, G., & Goldstein, T. (2017). Visualizing the Loss Landscape of Neural Nets. arXiv:1712.09913







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