# Parallelizing Optimization Algorithms for Machine Learning

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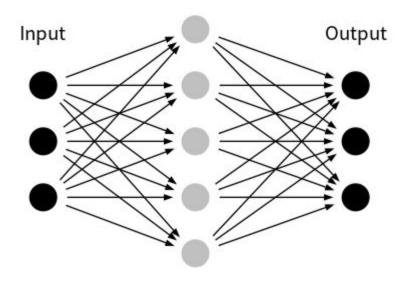
### Introduction

- Neural network implementation with C++
- Parallelize with OpenMP and CUDA
- MNIST
- Back propagation
- Stochastic gradient descent

#### Problem statement

- Optimization of neural network
- Parallelize back propagation
- Stochastic gradient descent

#### Hidden



## Proposed solution - Parallelization

- Linear layer
  - Matrix calculation
  - Vector addition
  - $\circ$  input  $\times$  Weight + Bias
- Activation Function (Sigmoid)
  - Elementwise operation
- Gradient
  - Independant calculation

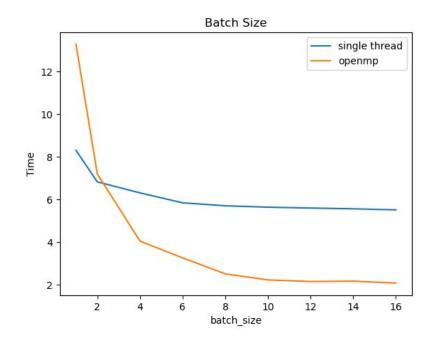
## Challenges encountered

- Memory management
- Race condition
- Runtime Error
- Implementation detail
- Hard to debug
- Compilation

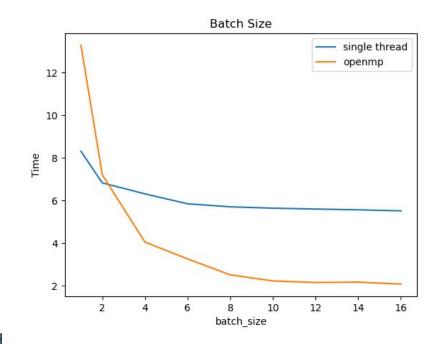
### **Evaluation Platform**

- i7-13700K (limited to 8 hardware threads)
- GeForce RTX 3090
- Ubuntu 22.04.1 LTS
- GCC 11.3.0
- CUDA 11.8

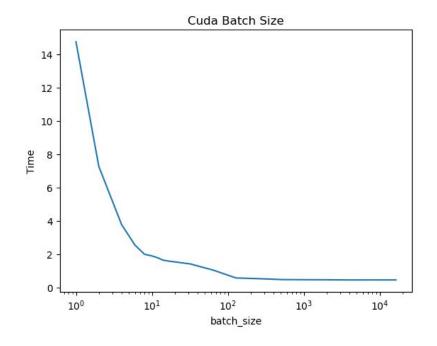
- Batch Size ↔ Time
  - Hidden Dimension = 300
- batch\_size is bigger
  - SGD iteration is smaller
  - zero\_gred, backward, update
  - single thread time get lower



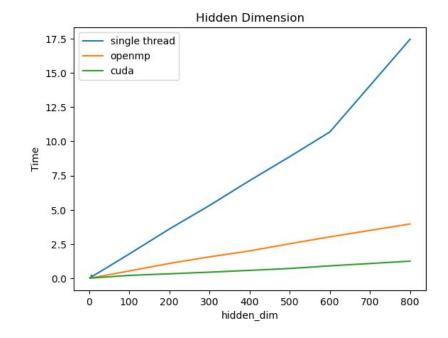
- Parallelized on batch\_size
  - o batch\_size 1 to 8
  - multi thread get parallelized
- Thread creation overhead
  - batch\_size 1 and 2
  - lower than single thread
- Limited to 8 hardware threads
  - batch\_size > 8, as single thread



- Parallelize with CUDA
- 10496 CUDA CORES in RTX-3090
- Parallelized on batch\_size
  - o batch\_size 1 to 1e4
  - multi thread get parallelized

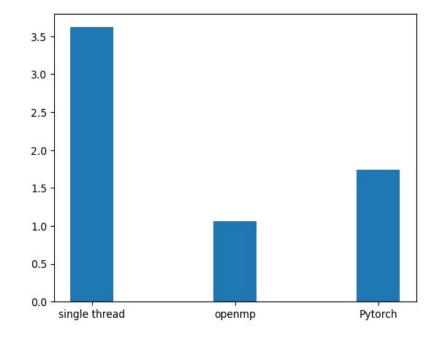


- Hidden Dimension ↔ Time
  - Batch size = 1024
- hidden\_dim of NN is bigger
  - computation get larger
  - time get longer for all cases



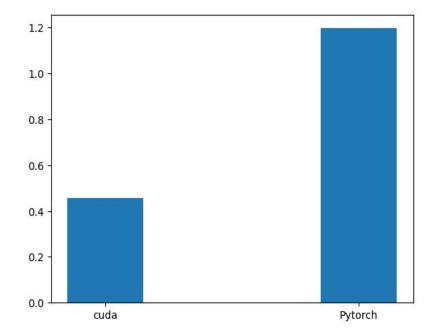
## Comparison - PyTorch

- CPU
- Parameters:
  - o batch\_size = 1024
  - o hidden\_dim = 300
- 1.7392157264985144 sec



## Comparison - PyTorch

- CUDA
- Parameters:
  - o batch\_size = 1024
  - o hidden\_dim = 300
- 1.1973519088700413 sec



#### Related work

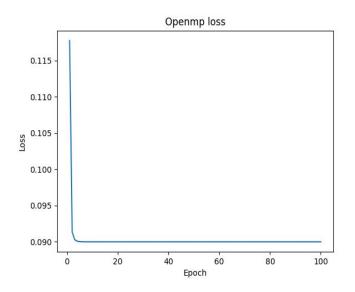
- PyTorch
- TensorFlow
- Keras

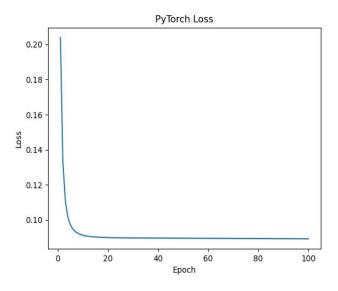






## Related work - PyTorch Loss





#### Contributions of each member

- 林律穎
  - Coding(20%), Script, Design Experiment, Running Experiment, Presentation
- 陳重諺
  - Coding(60%), Environment Setup, Design Experiment, Testing Platform 😍



- 林立倫
  - Coding(20%), Project Structure, Environment Setup, Presentation

#### Conclusion

- The degree of parallelism highly depend on batch size in our implementation
- Because in our implementation, we mostly use block partition on batch size.
- Block partition on batch size has less chance to create race condition.
- Put all data in device memory to lessen the overhead of memory transmission.
- And let managing memory become easier.
- Use libraries if there are existed ones

Q & A

## Thanks for listening