Training Deeper Models by GPU Memory Optimization on TensorFlow

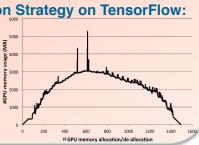


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Introduction

- 1."OOM" issue in training deeper Models.
- 2. The major constituents of memory usage is feature map.
- 3. Dynamic Allocation Strategy on TensorFlow:

Tensor will be released when reference count becomes 0.



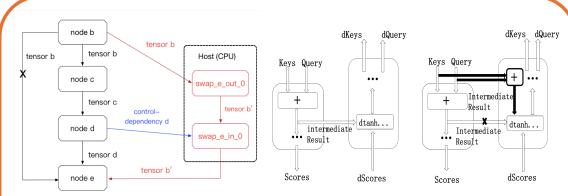


Figure 1. Atomic operation of the swap out/in optimization.

Figure 2. Optimization on Attention operation.

Our approaches

Swap out/in

> Rewrites the dataflow graph to utilize host memory as memory pool. (Fig. 1)

- Which feature maps to be swapped out?
- When to be swapped back in?

Memory-Efficient Attention

Attention layer actually requires much more memory space than LSTM/GRU layers in the Seg2Seg models. (Fig. 2)

- Drop Attention intermediate results directly.
- The re-computation cost is extremely cheap, just an addition operation.

Experiments and Results

Table 1: Evaluation of Swap out/in. GPU memory limit is 12GB

(a) General Models.			(b) ResNet.			
Model	B_{base}	B_{opt}	Model	M_{base}	M_{opt}	
ResNet-50	144	664(+361%)	ResNet-101	5815MB	2660MB	
Inception-V3	208	548(+163%)	ResNet-200	10662MB	3052MB	
GAN	24	48(+100%)	ResNet-1001	OOM	5979MB	
NMT	496	824(+66%)	ResNet-2000	OOM	10650MB	

Table 2: Evaluation of memory-efficient sequence models.

(a) TF-LM model.			(b) TF-NMT model.		
LSTM Layers	B_{base}	B_{opt}	Time Steps	B_{base}	B_{opt}
1	1800	3000(+67%)	50	350	1100(+214%)
4	750	1500(+100%)	100	90	550(+511%)
8	350	900(+157%)	200	20	230(+1050%)
16	75	280(+273%)	400	2	60(+2900%)

B_{oot}: max batch size after applying memory optimization. Mont: max memory usage after applying optimization

Conclusions

- The dataflow-graph based Swap out/in method.
- Memory Efficient Attention op to save huge amount of memory for Seq2Seq models.
- All approaches are integrated into TensorFlow seamlessly without requiring any changes to existing model descriptions.
- The max training batch size can be increased by 2 to 30 times.