• FP16 python ./main.py --mode=inference\_benchmark --use\_tf\_amp --warmup\_steps 20 --train\_iter 100 -iter\_unit batch --batch\_size <batch size> --data\_dir=<path to imagenet> --log\_dir=<path to results directory>

Each of these scripts, by default runs 20 warm-up iterations and measures the next 80 iterations.

To control warm-up and benchmark length, use --warmup\_steps, --num\_iter and --iter\_unit flags.

## Results

The following sections provide details on how we achieved our results in training accuracy, performance and inference performance.

## **Training accuracy results**

Our results were obtained by running the ./scripts/RN50\_{FP16, FP32}\_{1, 4, 8}GPU.sh script in the tensorflow-19.02-py3 Docker container on NVIDIA DGX-1 with 8 V100 16G GPUs.

number of GPUs	mixed precision top1	mixed precision training time	FP32 top1	FP32 training time
1	76.18	41.3h	76.38	89.4h
4	76.30	10.5h	76.30	22.4h
8	76.18	5.6h	76.26	11.5h

## **Training performance results**

Our results were obtained by running the  $\,$  ./scripts/benchmarking/DGX1V\_trainbench\_fp16.sh  $\,$  and

./scripts/benchmarking/DGX1V\_trainbench\_fp32.sh scripts in the tensorflow-19.02-py3 Docker container on NVIDIA DGX-1 with 8 V100 16G GPUs.

number of GPUs	mixed precision img/s	FP32 img/s	mixed precision speedup	mixed precision weak scaling	FP32 weak scaling
1	818.3	362.5	2.25	1.00	1.00
4	3276.6	1419.4	2.30	4.00	3.92
8	6508.4	2832.2	2.30	7.95	7.81

Our results were obtained by running the  $\label{local_problem}$  ./scripts/benchmarking/DGX1V\_inferbench\_fp16.sh and

## Inference performance results

batch size	mixed precision img/s	FP32 img/s
1	177.2	170.8
2	325.7	308.4
4	587.0	499.4
8	1002.9	688.3
16	1408.5	854.9

<sup>./</sup>scripts/benchmarking/DGX1V\_inferbench\_fp32.sh scripts in the tensorflow-19.02-py3 Docker container on NVIDIA DGX-1 with 8 V100 16G GPUs.