DEEP LEARNING BENCHMARKS ON
SUPERMICRO’S 4U 8 GPU SYSTEM BASED ON
DUAL 3RD GEN AMD EPYC™ PROCESSORS

TABLE OF CONTENTS

Executive Summary .........................1
System Configuration ......................2
Deep Learning Benchmarks ..............3
Deep Learning Benchmark Comparison
Using Different Workloads ...............4
Conclusion ................................5

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Demonstrating the performance of the Supermicro AS-4124GS-TNR, a
4U dual-processor 8 GPU server with up to 8TB of memory, and 160
Lanes of PCI-E 4.0, shows the generation over generation
performance improvements of the new 3rd Gen AMD EPYC 7003 Series
Processors on Deep Learning benchmarks

Executive Summary

Artificial Intelligence is being adopted in various industries worldwide. The
choice of systems to perform these complex tasks is critical and requires
understanding how the different system components act together. A series of
benchmarks have been created that allow those who evaluate systems and
architectures to determine which combination of CPUs and GPUs are the
best fit for their workloads.
AI workloads require optimized systems and need to incorporate the proper hardware and tuning the software to deliver maximum performance at a given price point. A solution that provides value to end-users consists of the choice of CPUs, GPUs, and the proper software stack. Various numbers of cores, communication latency between cores, GHz, and which generation of CPU architectures can influence benchmark performance of real-world AI applications.

A comparison will be run for this benchmark that compares 2nd Gen AMD EPYC™ processors to 3rd Gen AMD EPYC processors. AMD provides a wide range of processors with different numbers of cores and speed levels. Any AI/DL/ML application will depend heavily on the GPUs selected. Supermicro has run benchmarks that use different CPU generations and NVIDIA V100 and A100 GPUs. The CPU controls the management and assignment of work to the GPUs, while the GPU does the heavy lifting of transforming, loading, and analyzing the data. This is the training phase of AI deep learning, as well as inferencing.

**1. System Configuration**

Supermicro designs and delivers a wide variety of servers and storage systems to enterprises worldwide. For these benchmarks, the AS-4124GS-TNR was the system of choice and features dual AMD 2nd Gen or 3rd Gen EPYC processors, up to 160 PCI-E Gen 4 lanes and with up to 8 PCI-E GPUs.

The benchmarks that are used in this paper are widely available, as is the software stack. The table below describes the software stack used for these benchmarks and the URLs' location where the software can be downloaded. Table 1 and Table 2 list these components.

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<thead>
<tr>
<th>Software</th>
<th>Version</th>
<th>Source</th>
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<td>Docker</td>
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<tr>
<td>Mellanox driver</td>
<td>5.1</td>
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Table 1 - Software Specifications

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<th>Hardware</th>
<th>Items</th>
<th>Model</th>
<th>Description</th>
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<td>Supermicro GPU system SKU</td>
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<td>MB</td>
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<td>Motherboard BIOS 2.0</td>
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<td>CPU</td>
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<td>3rd Gen AMD EPYC CPU</td>
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<td>Memory</td>
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<td>NVIDIA A100 40GB PCIe Gen4 GPUS</td>
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<tr>
<td>Network</td>
<td>MT28800 ConnecX-5</td>
<td>Mellanox 100 Gbps AOC</td>
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</tr>
</tbody>
</table>

Deep Learning Benchmark

There are many ways to benchmark a GPU system with a Deep Learning workload. Many types of workloads can be run as benchmarks, and a comprehensive list, with details, methodologies, and required software components, is maintained on github.com.

Supermicro is the first to benchmark a system’s performance under different Neural Network applications, followed by benchmarking the GPU system with a real dataset. For comprehensive and a more controlled comparison of Deep Learning workloads, an increasing number of manufacturers and end customers are adopting the MLPerf suite, which covers a wide variety of AI/ML/DL workloads. Supermicro is committed to making the MLPerf benchmark as part of specifications for all GPU-capable systems.

The benchmarks that Supermicro ran are varied and need to be discussed separately.

1) Deep Learning performance with different Neural Network applications

Figure 1 shows the throughput of different Deep Learning Neural Network applications. The benchmark was run with the NVIDIA NGC container that the Deep Learning platform supplies, performs, and bare-metal systems. More details about Deep Learning Neural Network applications are available here for technical information about the Neural Network Applications.
To be compliant with NVIDIA NGC, Ready fundamental performance criteria, the Google Neural Machine Translation (GNMT), and ImageNet Classification are used in this benchmarking system.


There are many ways to benchmark a system in a given domain. Synthetic benchmarks are constructed to generate a specific workload on the underlying system and use its application to generate the data. Real-world workload benchmarks use actual data loaded into an application to produce results. Figure 2 and Figure 3 show the Deep Learning benchmark results with Synthetic and Real datasets, respectively.
To understand the basics of how this benchmark was set up and run, please look at the instructions posted [here](#). This document contains step-by-step procedures for using an NGC container benchmarking with the NVIDIA GPU system with both synthetic datasets and real datasets. There are also instructions for running the ResNet512 benchmark with TensorFlow. Visit this [page](#) to find out more about Deep Learning benchmark results and logs.

3) PyTorch GNMT Real Dataset Deep Learning performance with single-precision FP32

![TensorFlow ResNet50v1.50 Real Workload Benchmark ResNet50v1.50 FP32 ImageNet 2012 Dataset](#)

*Figure 3 - Results running the TensorFlow Benchmark on an AS-4124GS-TNR with 3rd Gen AMD EPYC 7313 CPUs*

![Google Neural Machine Translation Performance GNMT-WMNT16 Real Dataset FP32](#)

*Figure 4 - Results running the Google Neural Machine Translation Performance on an AS-4124GS-TNR with 3rd Gen AMD EPYC 7313 CPUs*
Deep Learning Benchmark Comparison using Different Workloads

Comparing benchmarking results with a similar base system and the same Deep Learning workload can give a business overview of how a GPU system throughput is improved by introducing more advanced CPUs and GPUs.

1) NVIDIA GPU – A100 vs. V100

The NVIDIA A100 GPU increases the Deep Learning training throughput and adds more features, such as TF32 Tensor Core and Multi-Instance-GPU (MIG). MIG virtualizes a physical GPU into seven different instances that isolate AI/ML/DL workloads and maximize GPU utilization without interference from another GPU. To learn more about TF32 and MIG, please read more at this link. Figure 5 shows the same Deep Learning workload on two different types of GPUs, and the throughput is up to 20-40%.

![Figure 5 - Comparison of NVIDIA A100 and V100 GPU Results](image)

2) AMD CPUs – 3rd Gen AMD EPYC (formerly codenamed "Milan") vs. 2nd Gen AMD EPYC (formerly codenamed "Rome")

Figure 6 indicates that the 3rd Gen AMD EPYC 7313 can perform better than the 2nd Gen AMD EPYC 7H12 on a comparable base system, training a model with ImageNet Synthetic dataset using single-precision FP32.

The 3rd Gen AMD EPYC 7313 has many new BIOS options critical to these new processors' high performance. As in previous generations of AMD EPYC processors, the setting of IOMMU and NPS are two of them that could significantly impact OS installation and overall performance. Please refer to the NVIDIA design guide, DG-10105-001, for PCIe servers. Tuning the CPUs to get datasets ready for Deep Learning applications is critical to both system designers and the end-user customers. Please look at the AMD resource guide for more information.
Benchmark results may vary if test conditions are different. The synthetic dataset indicates theoretical performance. Figure 6 is the benchmark results with the synthetic dataset. The real dataset, however, would more accurately represent Deep Learning application performance.

![NVIDIA A100 PCIe Performance](image)

*Figure 6 - Comparison of performance of 3rd Gen AMD EPYC vs. 2nd Gen AMD EPYC*

![Benchmark Results - Synthetic vs. Real](image)

*Figure 7 – Performance difference between Real and Synthetic Benchmarks with increasing number of GPUs.*
Figure 7 shows the benchmark results between the Synthetic and Real benchmarks with the ImageNet 2012 datasets on the 3rd Gen AMD EPYC CPU. It can be up to a 33% throughput difference between synthetic and real ImageNet 2012 dataset, using ResNet50 v1.50 in a multi-GPU system. Therefore, benchmarking a system with a real Deep Learning related dataset is essential to assess performance accurately.

Conclusion

The benchmark results clearly show that the AMD EYPC 7313 processors improve the NVIDIA GPU system’s throughput with Deep Learning workloads. A 15-30% generational increase in the synthetic benchmark test is seen with EPYC with the same NVIDIA A100 GPUs in a similar Supermicro chassis. The benchmark results also demonstrate that A100 PCIe can outperform V100 SXM2 in the comparable GPU systems up to 40%. Combined with the NVIDIA A100 GPU, Supermicro AMD CPU-based GPU systems are very flexible, competitive, and offer exceptional customer experiences. To learn more information on Supermicro GPU systems, please visit https://www.supermicro.com/en/products/GPU/

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