



# Accelerating Text-to-Image Diffusion Models using OpenVINO with Intel Xeon Processors

Planning / Implementation

Text-to-image diffusion models, such as Stable Diffusion XL, represent the frontier of generative AI, offering the ability to create realistic images directly from textual descriptions. These models, however, are computationally intensive and often require high-performance hardware typically found in GPU-based systems. This dependency on GPUs can present cost, availability, and power efficiency challenges—particularly at scale.

To address these limitations, this paper explores an alternative approach: accelerating text-to-image model inference on CPU-based infrastructure using Intel's OpenVINO toolkit and the Al acceleration capabilities of Intel Advanced Matrix Extensions (AMX), available in 5th and 6th Gen Intel Xeon Scalable processors.

Figure 1 illustrates sample images generated using OpenVINO with Stable Diffusion XL.



Figure 1: Example images generated using stabilityai's stable-diffusion-xl model with OpenVINO

This paper describes the following:

- Introduction to the Intel Xeon platform, explaining how AMX enhances deep learning performance
- A step-by-step walk-through of the implementation of Stable Diffusion XL using OpenVINO
- Performance benchmarking data comparing native Hugging Face inference with OpenVINOaccelerated pipelines

# Intel Xeon processors and Intel AMX

At the heart of this solution are Lenovo ThinkSystem servers with Intel Xeon processors. These processors include Intel Advanced Matrix Extensions (Intel AMX), which enable direct acceleration of deep learning inference and training workloads on the CPU.

By eliminating the dependency on discrete accelerators for many Al tasks, organizations benefit from simplified infrastructure, streamlined deployment, and a reduced total cost of ownership. With backward compatibility to prior generation Intel Xeon platforms, businesses can also preserve existing investments and accelerate time to value through seamless system upgrades.

In this paper we are using the ThinkSystem SR650 V3 as the basis for our testing. The SR650 V3 is based on 5th Gen Intel Xeon processors. These processors deliver up to:

- 14× PyTorch inference performance improvement over 3rd Gen Intel Xeon
- 5× better performance per watt compared to 4th Gen AMD EPYC
- Sub-100ms latency on large language models with up to 20B parameters

It is expected that you would get even greater results using the ThinkSystem SR650 V4 with the latest Intel Xeon 6700P processors.

## **Diffusion workflow**

This section outlines how to run an accelerated text-to-image diffusion model using OpenVINO on Intel CPUs.

The steps are as follows:

1. Create a virtual environment (Optional)

We recommend you install the uv package to create a virtual python environment to avoid dependency issues and quickly download needed python libraries. Otherwise, if you have a clean install or a containerized machine, you can install the needed libraries directly without a virtual environment.

Run the following in your Linux terminal:

```
curl -LsSf https://astral.sh/uv/install.sh | sh
uv venv diffusion-env
source diffusion-env/bin/activate
```

#### 2. Download dependencies

Install the necessary OpenVINO python libraries as well as the optimum command line tool to easily obtain pre-trained models from huggingface.

```
uv install optimum[openvino]
uv pip install -U --pre --extra-index-url https://storage.openvinotoolk
it.org/simple/wheels/nightly openvino openvino-tokenizers openvino-gena
i
```

#### 3. Download Models & Convert to OpenVINO-IR

Using the optimum command line interface download a huggingface model of your choosing and convert it to the OpenVINO Intermediate Representation.

Below is an example using the stable-diffusion-xl model due to its visually impressive results.

```
optimum-cli export openvino --model stabilityai/stable-diffusion-xl-bas e-1.0 /~ /.cache/huggingface/hub/stable-diffusion-xl-base-1.0-ov
```

## 4. Create Inference Pipeline and Generate Image

Begin using the model to generate a sample image using the following python script. The results should be a 1024x1024 image similar to Figure 2.

```
model_path = "/home/lenovoai/.cache/huggingface/hub/stable-diffusion-xl
-base-1.0-ov/" prompt = "A fox wearing a detective trench coat solving
a mystery"
pipe = ov_genai.Text2ImagePipeline(model_path, "CPU")
image_tensor = pipe.generate(prompt, num_inference_steps=40, rng_seed=4
2)

# Post-Process and Save Image
image = Image.fromarray(image_tensor.data[0]) image.save("output-img.pn
g")
```



Figure 2: Example output from prompt: "A fox wearing a detective trench coat solving a mystery"

# **OpenVINO performance**

To assess the performance benefits of OpenVision and Intel Xeon processor-optimized inference, a benchmark was conducted comparing Stable Diffusion XL-1.0 executed with native Hugging Face libraries with the OpenVINO toolkit on 5th Gen Intel Xeon processors. The results are shown in Figure 3.

Stable Diffusion XL-1.0 was selected due to its computational intensity and model complexity, representing a demanding use case for generative AI workloads. The test involved generating 30 images, each using 40 inference steps.

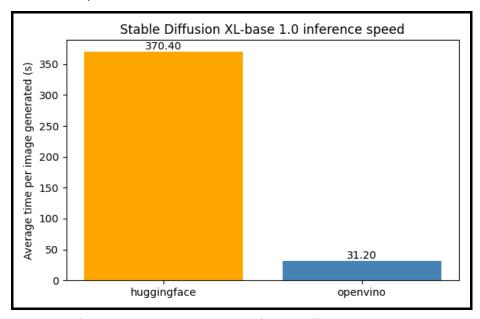


Figure 3. Inference speed comparison on Stable Diffusion XL-1.0 averaged over 30 images with 40 inference steps each; see Hardware Details for exact hardware specifications

Results demonstrate that native Hugging Face pipelines are not fully optimized to take advantage of the Al acceleration capabilities of Intel Advanced Matrix Extensions (Intel AMX) as implemented in 5th Gen Intel Xeon processors. Using the native implementation, the average time to generate a single image was approximately six minutes.

In contrast, OpenVINO significantly improves inference speed by applying graph-level optimizations and utilizing low-level hardware acceleration resulting in the average to perform the same task of thirty seconds, a 12x performance improvement. These enhancements enable more efficient execution of compute-intensive diffusion models on CPU-based infrastructure, delivering faster time-to-result and improving system throughput for Al-driven image generation workloads. This performance uplift makes CPU-based inference a viable, cost-effective alternative for organizations deploying large-scale generative Al applications.

#### Conclusion

This paper highlights a powerful, cost-effective solution for Generative AI: high-performance text-to-image inference on CPU-based infrastructure using Lenovo ThinkSystem servers with Intel Xeon Scalable processors and the OpenVINO toolkit. By harnessing Intel AMX acceleration in 5th Gen Intel Xeon Processors and OpenVINO's optimized model execution, organizations can achieve up to 12× faster inference over native libraries—without relying on discrete GPUs.

The result is a highly scalable, energy-efficient solution that reduces total cost of ownership, simplifies deployment, and supports sustainable AI operations. For enterprises looking to scale generative AI across production environments, Lenovo and Intel deliver a robust, ready-to-deploy platform that meets today's performance, efficiency, and operational demands.

The latest Intel Xeon 6 processors expand on the capabilities of their predecessors by offering broader support for Intel AMX, higher core counts, and improved power efficiency. Users can expect even faster inference times, greater throughput, and better energy-per-inference metrics. These enhancements further close the performance gap between CPU and GPU solutions while maintaining the flexibility and scalability of general-purpose server infrastructure. This makes Intel Xeon 6 an ideal foundation for next-generation generative AI deployments across enterprise, edge, and cloud environments.

Ready to accelerate your AI strategy? Explore how Lenovo and Intel can help your organization deploy high-performance, efficient, and reliable generative AI solutions at scale.

## **Hardware Details**

The following table lists the key components of the server we used in our performance tests.

Table 1. Hardware Details

| Component        | Description                                            |
|------------------|--------------------------------------------------------|
| Server           | Lenovo ThinkSystem SR650 V3                            |
| Processor        | 2x Intel Xeon Platinum 8592+ CPU @ 2.50GHz             |
| Installed Memory | 16x 32GB TruDDR5 4800MHz (2Rx8) RDIMM                  |
| Disk             | 1x 7450 960GB M.2 NVMe SSD, 1x 2.0TB M.2 NVMe SSD      |
| OS               | Ubuntu 22.04.5 LTS (GNU/Linux 6.8.0-60-generic x86_64) |
| OpenVINO         | 2025.2.0                                               |

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## Related product families

Product families related to this document are the following:

- Artificial Intelligence
- Processors

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