

# GPU Training Speed Tiers (2026)

A quick, practical ranking for deep learning training throughput

**PSA:** Mind you, this list was **generated by Gemini 3 Pro** as a quick guide. Although the few comparisons I have checked so far hold when looking at serious testing sites, do check actual speeds yourselves before making any relevant decision. You can check sites like [trooper.ai/benchmarks](https://trooper.ai/benchmarks)

Be aware of how large the VRAM of each GPU is, as this could break your code if you pretend to load a model + batch\_size bigger than the VRAM available for the GPU that you choose.

## Tier 1: The Hyperspeed Class

(100% – 90% Speed)

The fastest silicon in existence. These cards chew through gradients.

### 1. NVIDIA RTX 5090 (32GB)

**Speed Score:** 100/100

**Why:** The Blackwell architecture + GDDR7 memory makes this significantly faster than the A100 in raw floating-point operations (FLOPS). If the batch fits, nothing beats this card in 2026.

### 2. NVIDIA RTX 4090 (24GB) / NVIDIA A100 (40GB/80GB)

**Speed Score:** ~85–90/100

**The Comparison:**

- **RTX 4090:** Actually faster than the A100 in raw compute (TFLOPS) for FP16/BF16 training on smaller models.
- **A100 (Colab Pro):** Wins on memory bandwidth (HBM2e). It feeds data to the chip faster. In practice, they are often neck-and-neck for training speed, with the A100 being more consistent for massive batches.

## Tier 2: The High-Performance Class

(75% – 60% Speed)

Excellent for serious Deep Learning work.

### 3. NVIDIA RTX 5080 (16GB)

**Speed Score:** ~75/100

**Why:** Faster cores than the 3090/4080, but the memory bus is narrower than the 4090. It has the raw horsepower of a Ferrari but the gas tank (VRAM) of a sedan.

### 4. NVIDIA RTX 4080 Super / RTX 4080 (16GB)

**Speed Score:** ~65/100

**Why:** A very fast card, significantly ahead of the 3090 in pure compute, though it lacks the memory bandwidth of the top tier.

### 5. NVIDIA RTX 3090 Ti / RTX 3090 (24GB)

**Speed Score:** ~60/100

**Why:** The old king. It relies on brute force and massive memory bandwidth (936 GB/s). It is roughly equal to the RTX 5070 Ti in speed but holds its ground due to that wide memory bus.

### 6. NVIDIA V100 (16GB/32GB)

**Speed Score:** ~58/100

**Why:** Found in older Colab Pro instances. It uses HBM2 memory (super fast), so for certain matrix multiplications, it can still beat modern consumer cards like the 4070 Ti.

### Tier 3: The Mid-Range Workhorses

(50% – 35% Speed)

Great for fine-tuning standard models (BERT, ResNet, Llama-7B).

#### 7. NVIDIA RTX 5070 Ti / RTX 4070 Ti Super

**Speed Score:** ~50/100

**Why:** Fast modern cores, but narrow memory buses limit them compared to the 3090/4080 class.

#### 8. NVIDIA RTX 3080 Ti / RTX 3080 (10GB/12GB)

**Speed Score:** ~45/100

**Why:** Still a beast. Its memory bandwidth (760 GB/s) is actually higher than the RTX 4070 series, making it surprisingly fast for training even in 2026.

#### 9. NVIDIA L4 (24GB)

**Speed Score:** ~42/100

**Why:** The standard "modern" Colab GPU. It is roughly equivalent to an RTX 3070 or 4070 in speed, but optimized for enterprise stability.

#### 10. Apple M3 Ultra (Mac Studio)

**Speed Score:** ~40/100

**Why:** Warning: Benchmarking Apple vs. NVIDIA is tricky. In raw training (using MPS), the M3 Ultra behaves roughly like an RTX 3080 or 4070. It is much slower than a 4090, but it has the memory capacity of a server farm.

#### 11. NVIDIA RTX 5070 / RTX 4070

**Speed Score:** ~38/100

**Why:** Very efficient, but strictly middle-of-the-road for training speeds.

### Tier 4: The Entry Level

(30% – 15% Speed)

Functional, but you will be waiting a while.

#### 12. Apple M4 Max / M3 Max

**Speed Score:** ~30/100

**Why:** Roughly comparable to an RTX 3070 / 4060 Ti in training speed. Incredible for a laptop, but not a dedicated training monster.

#### 13. NVIDIA RTX 4060 Ti (16GB) / RTX 3070 Ti / RTX 3070

**Speed Score:** ~25/100

**Why:** The 4060 Ti is popular for its 16GB VRAM, but its raw compute speed is quite low due to a tiny 128-bit memory bus. It is slower than the 3070 in many tasks.

#### 14. Apple M2 Ultra

**Speed Score:** ~25/100

**Why:** Slower cores than the M3/M4, but massive bandwidth. Comparable to a 3070 Ti.

#### 15. NVIDIA RTX 3060 Ti / RTX 4060 / RTX 3060 (12GB)

**Speed Score:** ~20/100

**Why:** The baseline. The 3060 (12GB) is beloved not for speed, but for fitting models. It is slow.

## Tier 5: The "Slow Lane"

(*<15% Speed*)

Patience is required.

### 16. NVIDIA Tesla T4 (Colab Free Tier)

**Speed Score:** ~15/100

**Why:** It uses the Turing architecture (2018). It is significantly slower than even a basic RTX 3060. It takes about 3–4x longer to train a model on a T4 than on a 3080.

### 17. Apple M4 Pro / M3 Pro / M2 Max

**Speed Score:** ~12/100

**Why:** Good for inference, weak for training.

### 18. Apple M1 Ultra

**Speed Score:** ~10/100

**Why:** First-gen silicon. It shows its age now.

### 19. Apple M4 / M3 / M2 / M1 (Base Chips)

**Speed Score:** <5/100

**Why:** These have no active cooling (MacBook Air) or very few GPU cores. Do not train on these.