# METODE INTELIGENTE DE REZOLVARE A PROBLEMELOR REALE

GPUs details Laura Dioşan

## Costs for training DL models

Amortized hardware and energy cost to train frontier AI models over FPOCH AI time



Check more details here <u>How Much Does It Cost to Train Frontier</u> <u>AI Models? | Epoch AI</u>

#### **GPUs**

FLOP = "Floating Point Operation"; one addition, multiplication, etc TFLOP = 1 trillion FLOPs  $(10^{12})$ 

A100

Memory

Capacity 40GB HBM2 Bandwidth 1.5 TB/sec

Compute

FP64 9.7 TFLOPS/sec FP32 19.5 TFLOPS/sec BF16 39 FLOPS/sec FP1678 TFLOPS/sec

Tensor Cores

TF32: 156 TFLOPS/sec

FP16/BF16: 312 TFLOPS/sec

L40S

Memory

Capacity 40GB HBM2 Bandwidth 1.6 TB/sec

Compute

FP64 not supported FP32 90 TFLOPS/sec BF16 700 TFLOPS/sec FP16 700 TFLOPS/sec

**Tensor Cores** 

TF32: 366 TFLOPS/sec

FP16/BF16: 733 TFLOPS/sec

H100

P

Memory

Capacity 40/80GB HBM2 Bandwidth 3.0 TB/sec

Compute

FP64 30 TFLOPS/sec FP32 60 TFLOPS/sec BF16 120 TFLOPS/sec FP16 120 TFLOPS/sec

**Tensor Cores** 

TF32: 500 TFLOPS/sec

FP16/BF16: 1000 TFLOPS/sec

H200

Memory

Capacity 80GB HBM2 Bandwidth 2 TB/sec

Compute

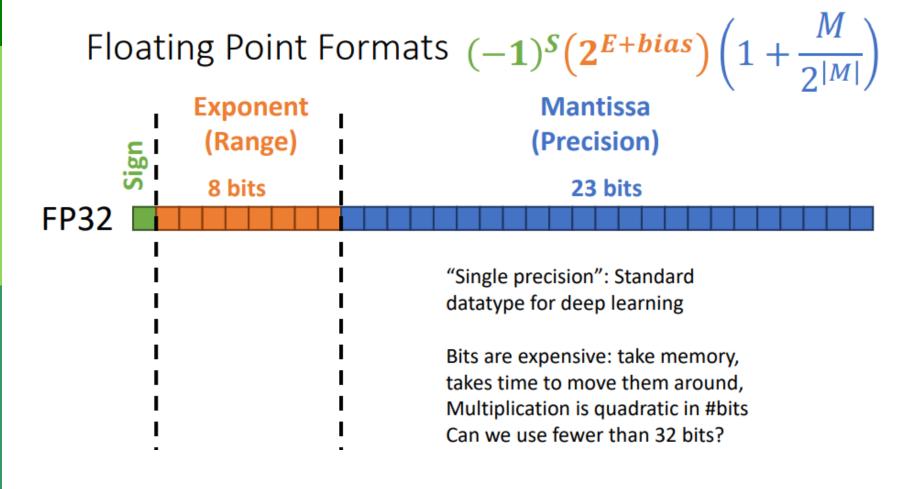
FP64 34 TFLOPS/sec FP32 67 TFLOPS/sec BF16 2000 TFLOPS/sec FP16 2000 TFLOPS/sec

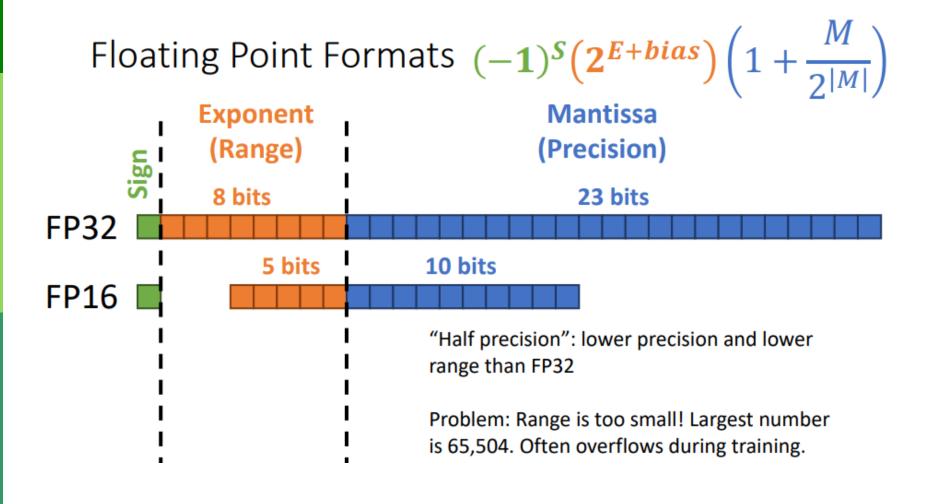
**Tensor Cores** 

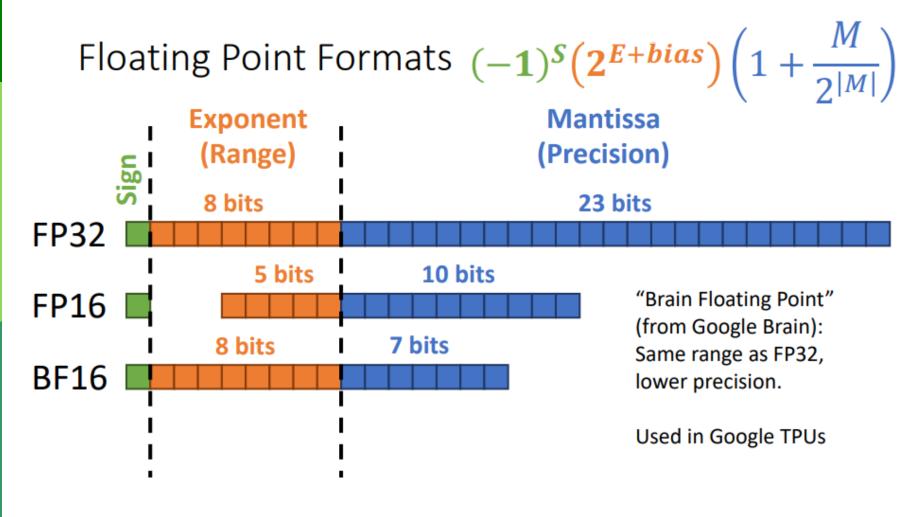
TF32: 989 TFLOPS/sec

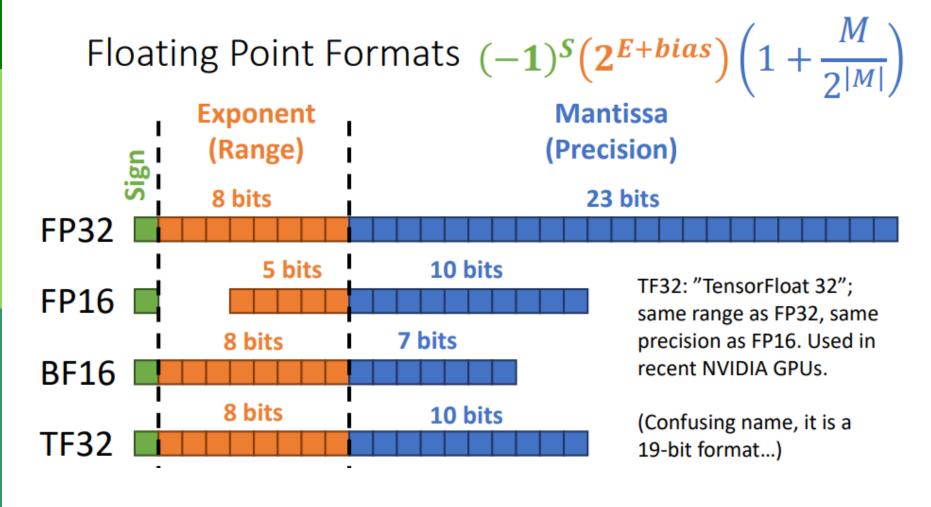
FP16/BF16: 1979 TFLOPS/sec











#### Mixed Precision

We often need to compute dot products (for matrix multiply, convolution, etc):

$$y = x_1 w_1 + x_2 w_2 + \dots + x_n w_n$$

Multiplication is more expensive than addition

**Idea**: Multiply in low precision, add in high precision

**Inputs**:  $x_i$ ,  $w_i$  in low precision (FP16, BF16, TF32)

**Output**: *y* in high precision (FP32)

$$y = FP32(x_1w_1) + FP32(x_2w_2) + \dots + FP32(x_nw_n)$$

**Tensor Cores** in NVIDIA GPUs are special hardware for mixed-precision matrix multiplication with different low-precision formats (TF32, BF16 best for neural nets)