

# Distributed Training

02457 Machine Learning Operations
Nicki Skafte Detlefsen,
Postdoc
DTU Compute

### What is distributed computations?



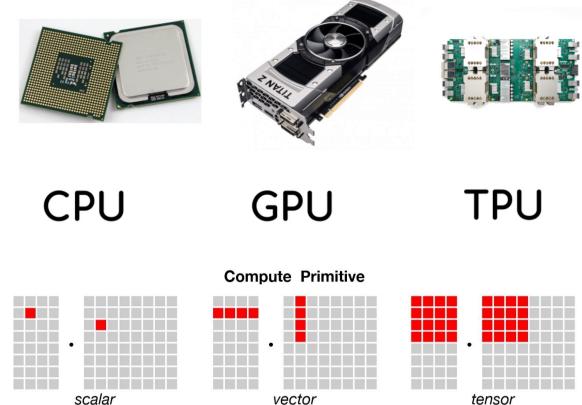
We focus on training as it is the most computationally expensive part of MLOps

#### Devices



- Three types of devices
  - CPU
    - General compute unit
    - 2-128 threads
  - GPU
    - Rendering unit
    - 1000-10000 threads
  - TPU
    - Specialized unit
    - 8-2048 threads

Note that we are comparing apples to bananas!



#### Memory

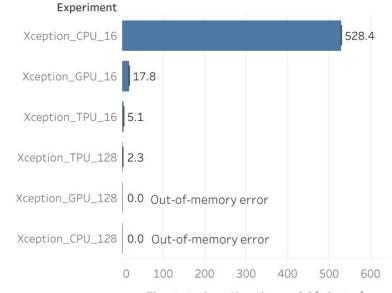


# Equally important to what device you are using, is the amount of memory that you have available

#### With more memory

- Faster data transfer
- Higher data modality
- Larger models

	CPU	GPU	TPU
Standard	32-64 GiB	12 GiB	64 GiB
Maximum	256 GiB	24 GiB	32 TiB



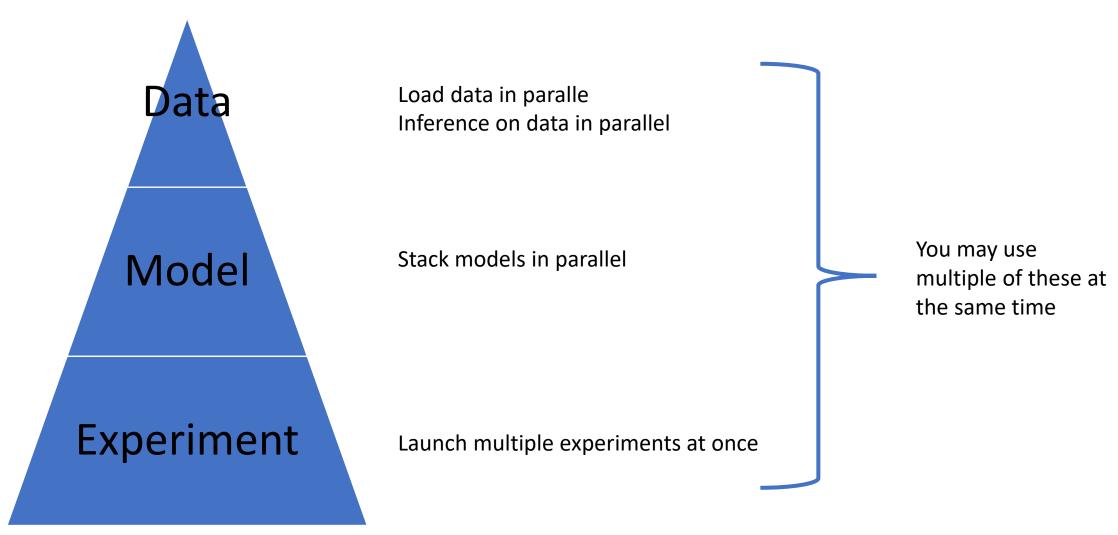
Time to train an Xception model (minutes)

Figure 3: CPUs vs GPUs vs TPUs for training an Xception model for 12 epochs. Y-Axis labels indicate the choice of model, hardware, and batch size for each experiment. Increasing the batch size to 128 for TPUs resulted in an additional ~2x speedup.

https://towardsdatascience.com/whento-use-cpus-vs-gpus-vs-tpus-in-a-kagglecompetition-9af708a8c3eb

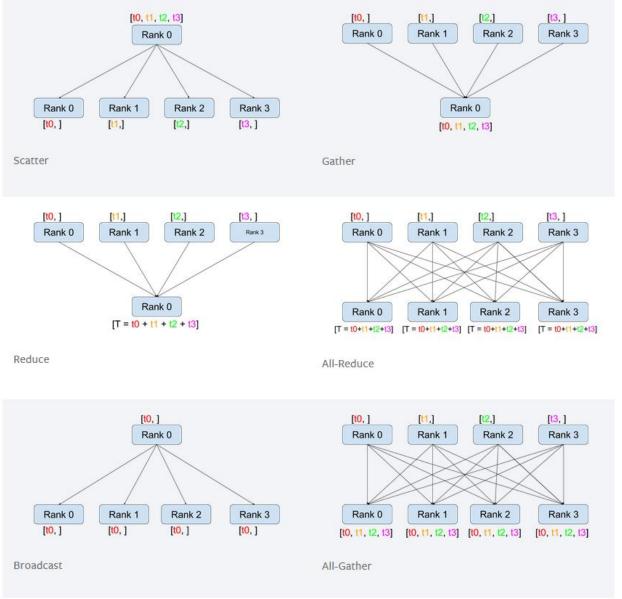
### Many layers of distributed computations





### The six imporatnt communication types

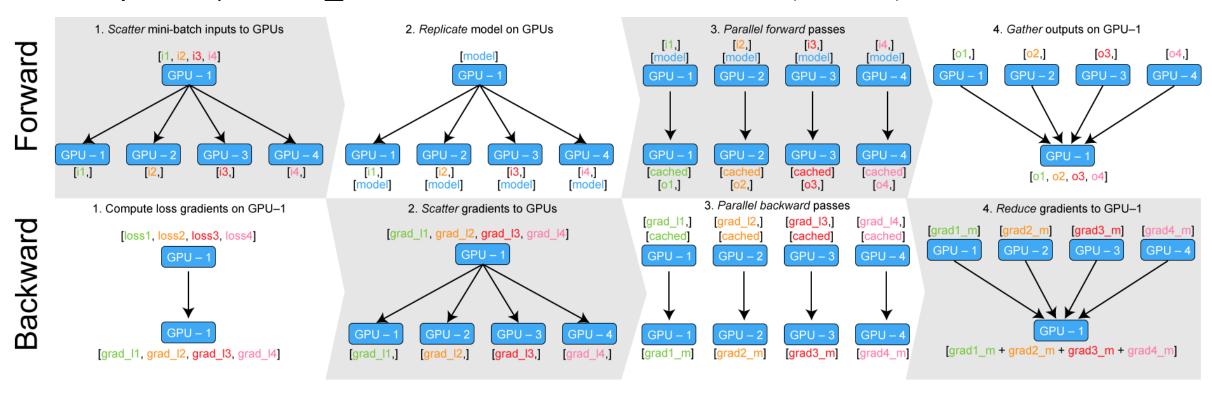




#### Data Parallel



#### Simple as parallel\_model = torch.nn.DataParallel(model)



## Why using a training framework



- Reduce boilerplate = increase turn-around time
- Focus on what is important
- Reproduceability
- Shareability
- Consistency
- Scaleability

### Training Fremworks





Many fremeworks exist for reducing boilerplate







Many frameworks for accelerating training







## Seperating engineering and research code

