



A Deep Learning tool for fast simulation

HNSciCloud Pilot Phase Final Public session

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Deep Learning for fast detector simulation

Detailed simulation of subatomic particles interactions is essential

Monte Carlo approach is not fast enough for HL-LHC needs

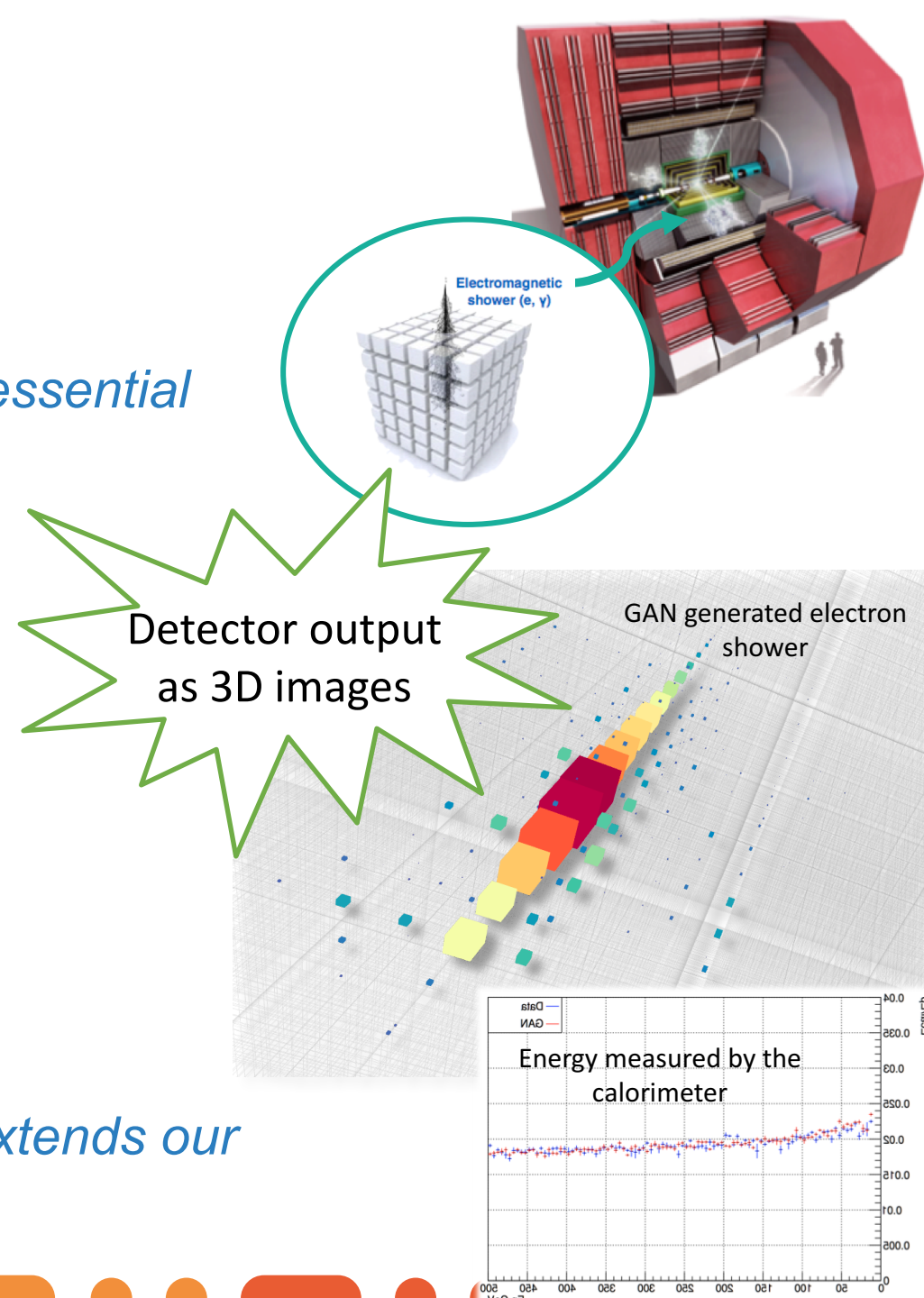
3D convolutional GAN generate realistic detector output

>2000x faster than Monte Carlo on a Intel Xeon processor

Training takes ~1 day on NVIDIA P100

Use parallel approach to distribute training across multiple nodes

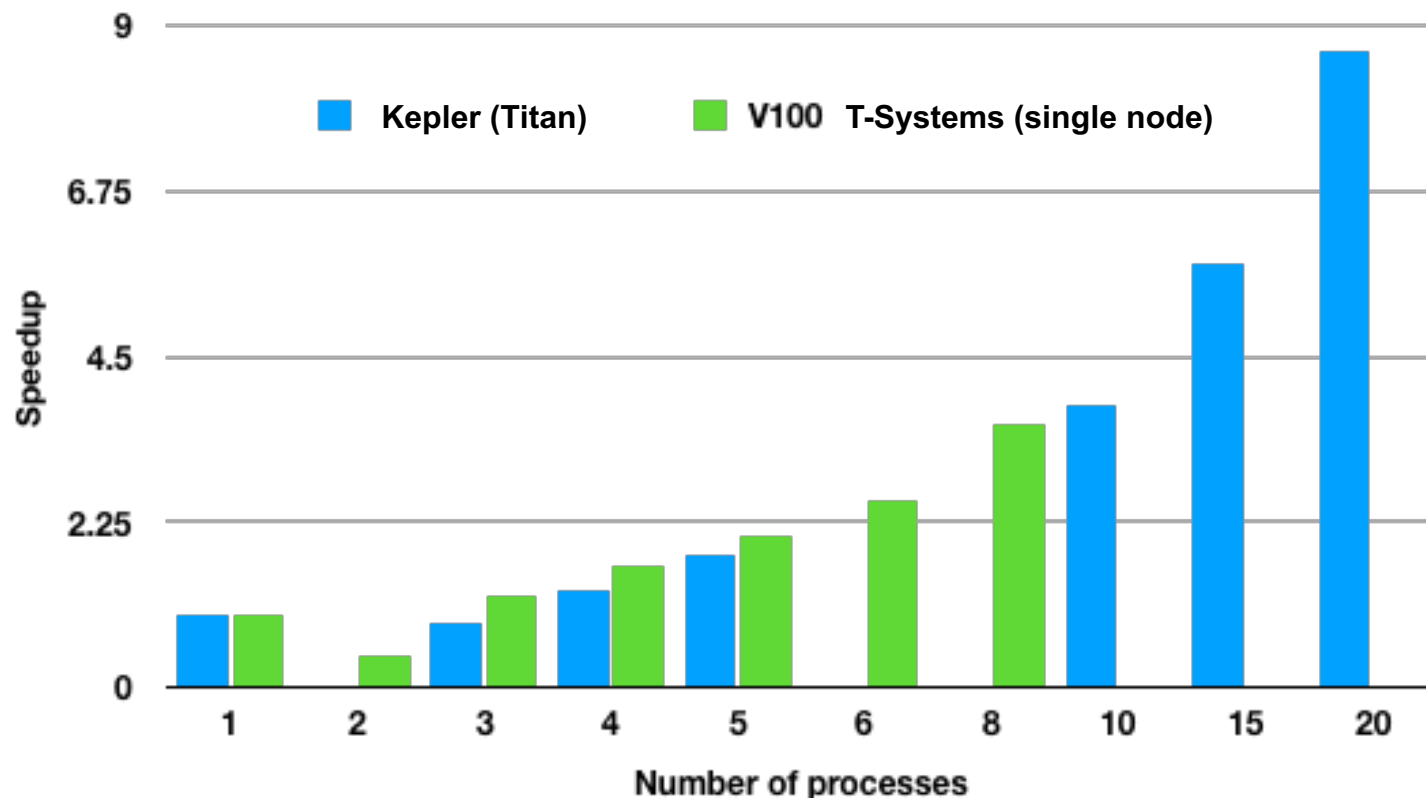
Optimized training and access to cloud facilities extends our range of applications



Distributed training on T-Systems cloud

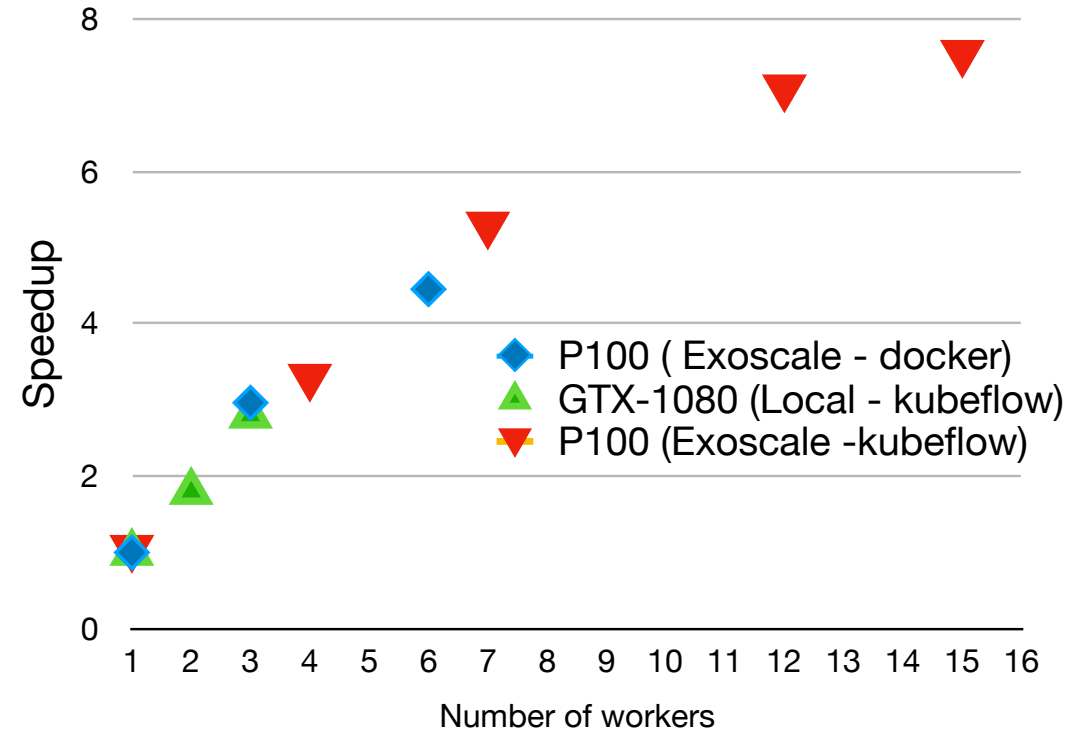
Introduce mpi based data parallel training

- First results on 8x NVIDIA V100 (BMS)
- 8 processes run in parallel on the same BMS
 - Better scaling than HPC
- Added values:
 - Access to powerful cluster
 - Simple and efficient instance creation



Distributed training on Exoscale cloud

- Failed attempts:
 - docker + HTCondor
 - Nuvla SLURM-based script
- Move to docker + Kubernetes/Kubeflow (K. Samaras-Tsakiris and R. Brito da Rocha from CERN IT-CM)
 - First results on 18x NVIDIA P100 (VMs)
 - Direct comparison to on-site GPUs in openStack
- Added values:
 - Large3 number of VMs
 - S3 storage solution available. Possibility to define shared FS across VMs?





Thanks!

Questions?

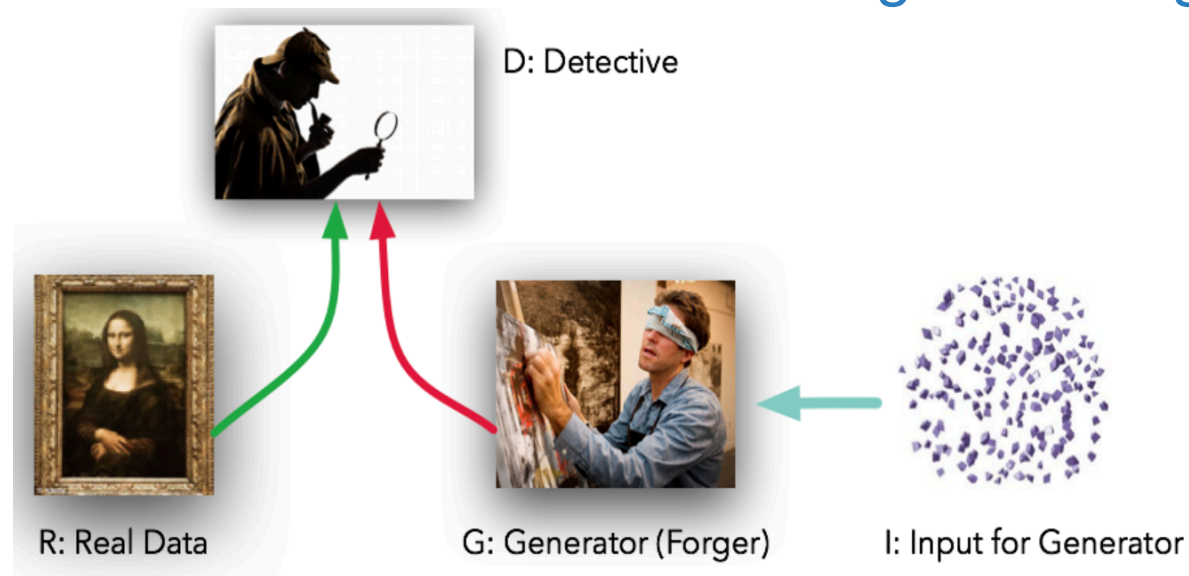
Generative adversarial networks

Generative methods from computer vision generate detector output!

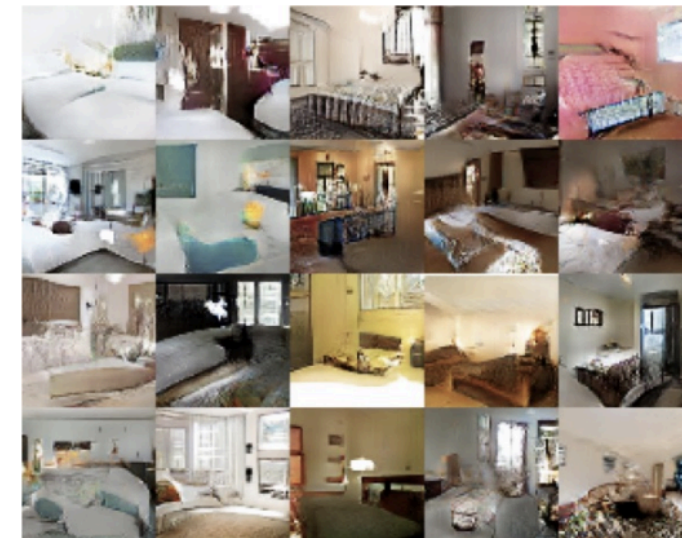
Simultaneously train two networks that compete and cooperate with each other:

Generator G generates data from random noise

Discriminator D learns to distinguish real/generated data



[Image source:](#)



<https://arxiv.org/pdf/1701.00160v1.pdf>

The counterfeiter/detective case

Counterfeiter shows the Monalisa

Detective says it is fake and gives feedback

Counterfeiter makes new Monalisa based on feedback

Iterate until detective is fooled