

Defense Data Generation in Distributed Deep Learning System

Se-Yoon Oh / ADD-IDAR

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syoh@add.re.kr

Overview

1. Introduction

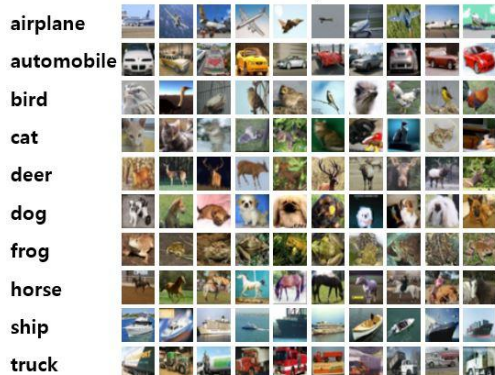
2. Data Generation Synthesis

3. Distributed Deep Learning

4. Conclusions

• ML Process (Supervised Learning, DNN)

① Labeled Training Images



Neural Network Framework



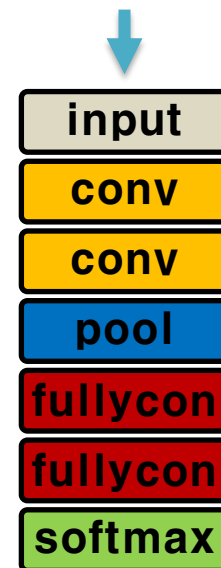
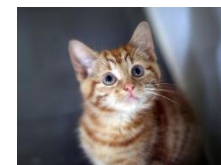
③ NN Model Definition

```
model.add(Convolution2D(128, 3, 3))
```

Train Model on GPU Accelerated Cluster



Trained Network



CAT

• ML Process (Supervised Learning, DNN)

① Labeled Training Images



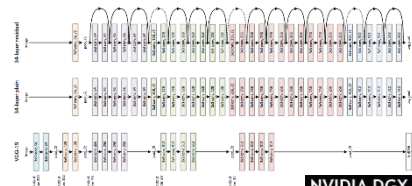
Benchmark DB ?
Transfer learning ?
One-shot learning ?

② Neural Network Framework



③ NN Model Definition

```
model.add(Convolution2D(128, 3, 3))
```



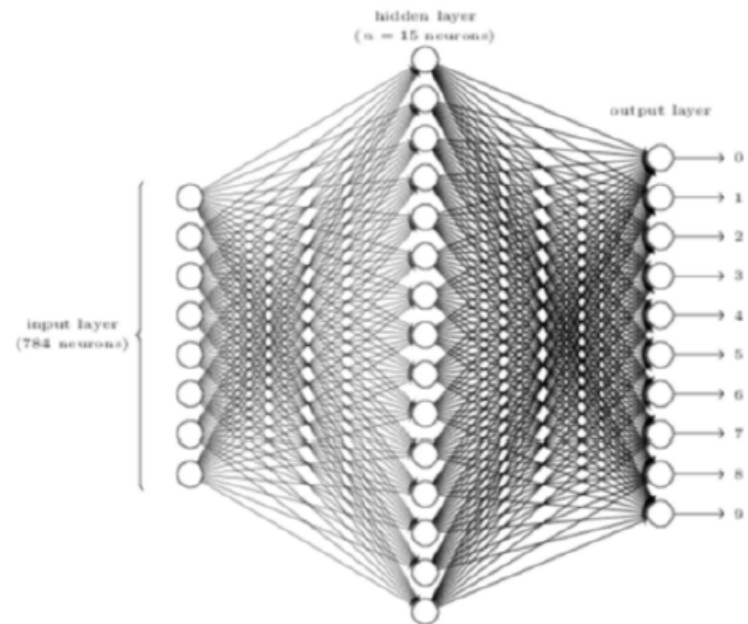
④ Train Model on GPU Accelerated Cluster



• NN Hyper-Parameter

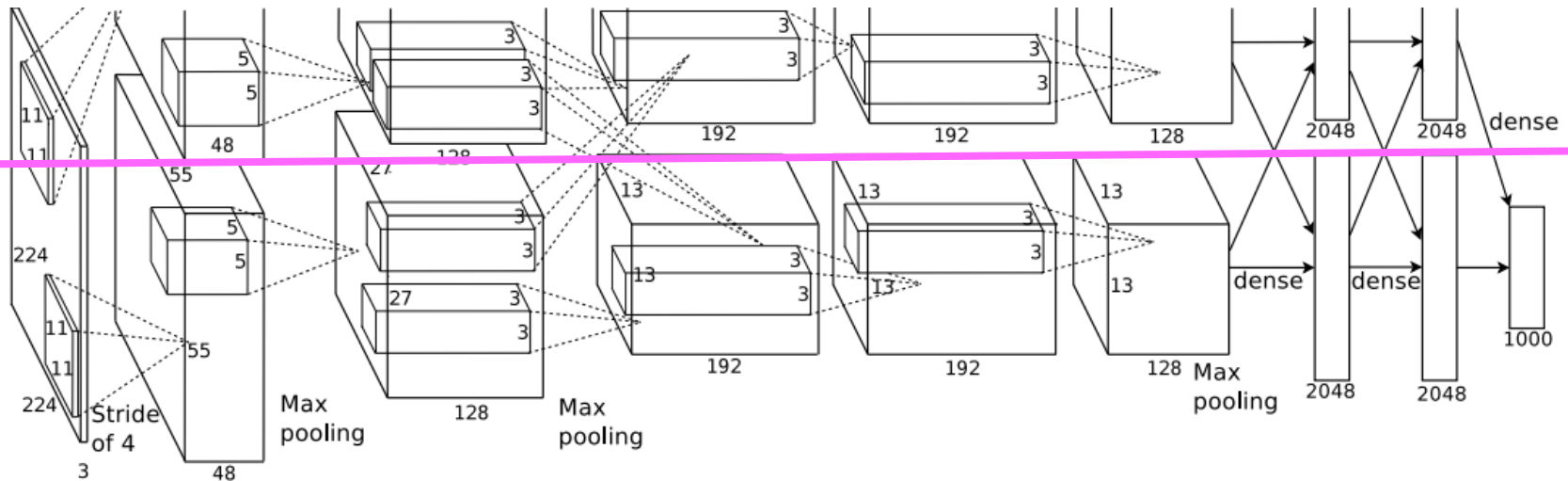
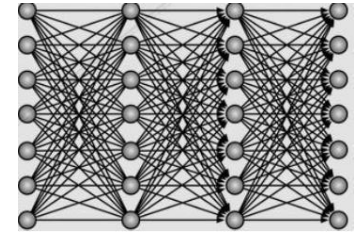
Hyper-Parameters

- Learning rates
- Cost function
- Regularization parameter
- Batch size
- Training epochs
- No. of layers
- Convolutional filters
- Convolutional kernel size
- Weight initialization
- Dropout fractions
- ...



ImageNet Classification with Deep Convolutional Neural Networks (Alex Krizhevsky, 2012)

- 650,000 neurons, 60,000,000 parameters
- 1.2 M images(224*224*3), 1,000 classes
- 7 Hidden layers, ReLU, DropOut, 2 GPU(GTX580)/week
- Overlapped max pooling, Augmentation



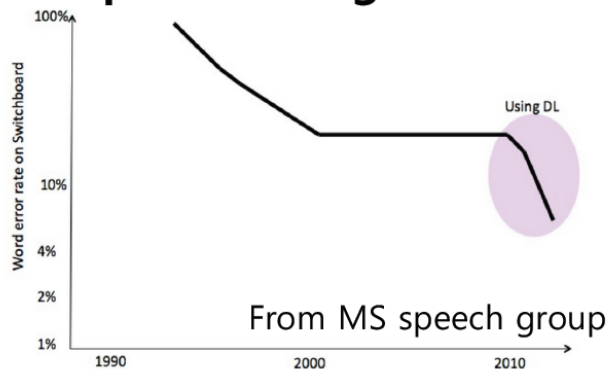
1st Layer

- 290,400개 (N)
- $((11 \times 11 \times 3) + 1) \times 96 = 34,944$ 개 (P)
- $290,400 \times 364 = 105,750,600$ 개 (C)

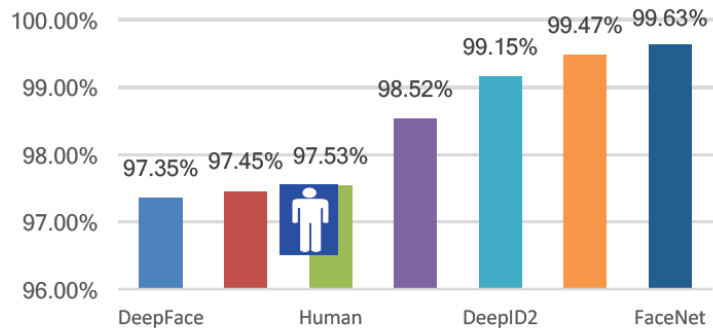
ConV L : 90~95% 연산량 (5% Para)
Full L : 5~10% 연산량 (95% Para)

• Tech. Trends

Speech recognition



Face verification (LFW)



Object Classification(ImageNet)

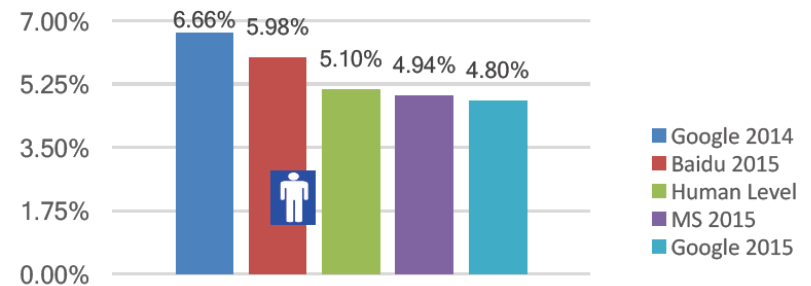


Table 3. ILSVRC classification task top-5 performance (with provided training data only).

Team	Time	Place	Top-5 error
SuperVision	2012	1	16.42%
ISI	2012	2	26.17%
VGG	2012	3	26.98%
Clarifai	2013	1	11.74%
NUS	2013	2	12.95%
ZF	2013	3	13.51%
GoogLeNet	2014	1	6.66%
VGG ³	2014	2	7.32%
MSRA	2014	3	8.06%
Andrew Howard	2014	4	8.11%
DeeperVision	2014	5	9.51%
MSRA PReLU-nets	2015.2	-	4.94%
BN-Inception	2015.2	-	4.82%
Deep Image	2015.5	-	4.58%



-
- **Supervised learning(SL)**
 - Standard in the current state of ML
 - Performed well on many task
 - Ground-truth annotation/label is **required** in training
 - **Unsupervised learning(UL)**
 - How most learning is done in the **real** world
 - **Impractical** to accurately annotate data
 - UL seems to be the key to **real** artificial intelligence
 - **Transfer learning & fine-tuning**
 - CNN -> Training with millions of unlabeled images(UL) -> **Deep understanding** in images
 - Quickly learn the new task with **much smaller labelled images (...?)**

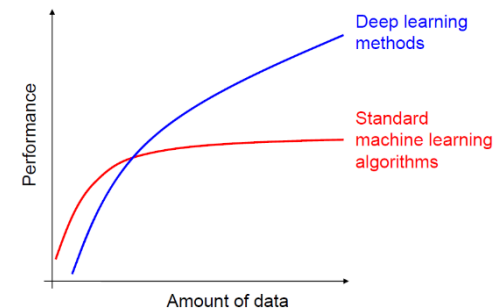
- **Two Things Considered**

- **Labeled data**

- Few labeled data in defense
 - Classification/object detection/segmentation/tracking/autonomy/...

- **Computing power**

- ML/DL based on HPC & GPU
 - Big data hub & data analytics



Overview

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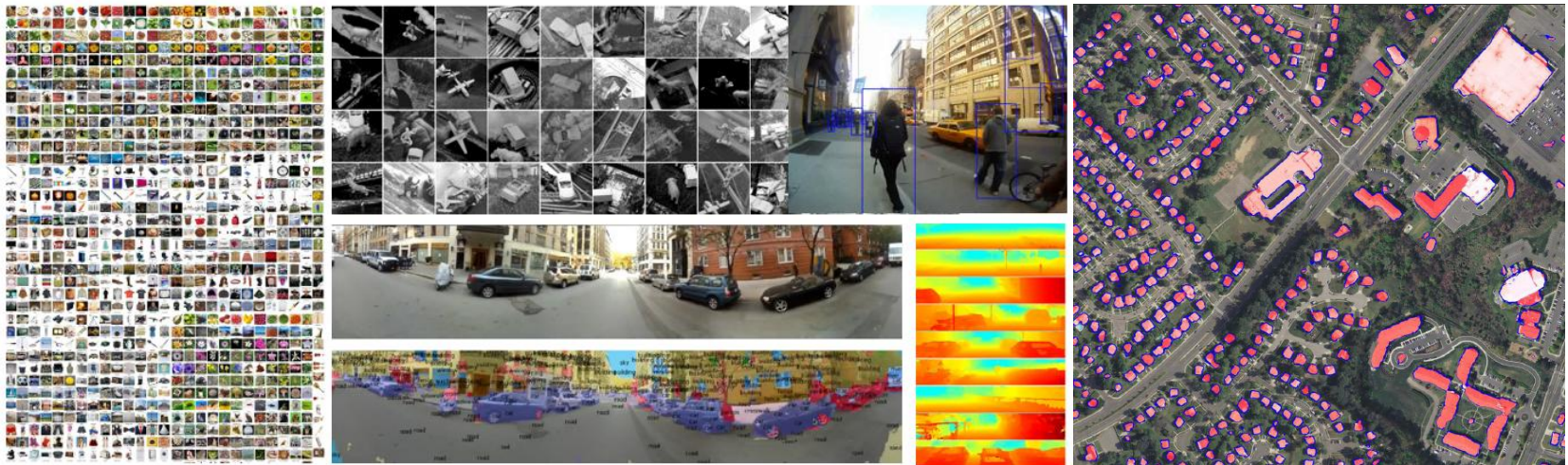
2. Data Generation Synthesis

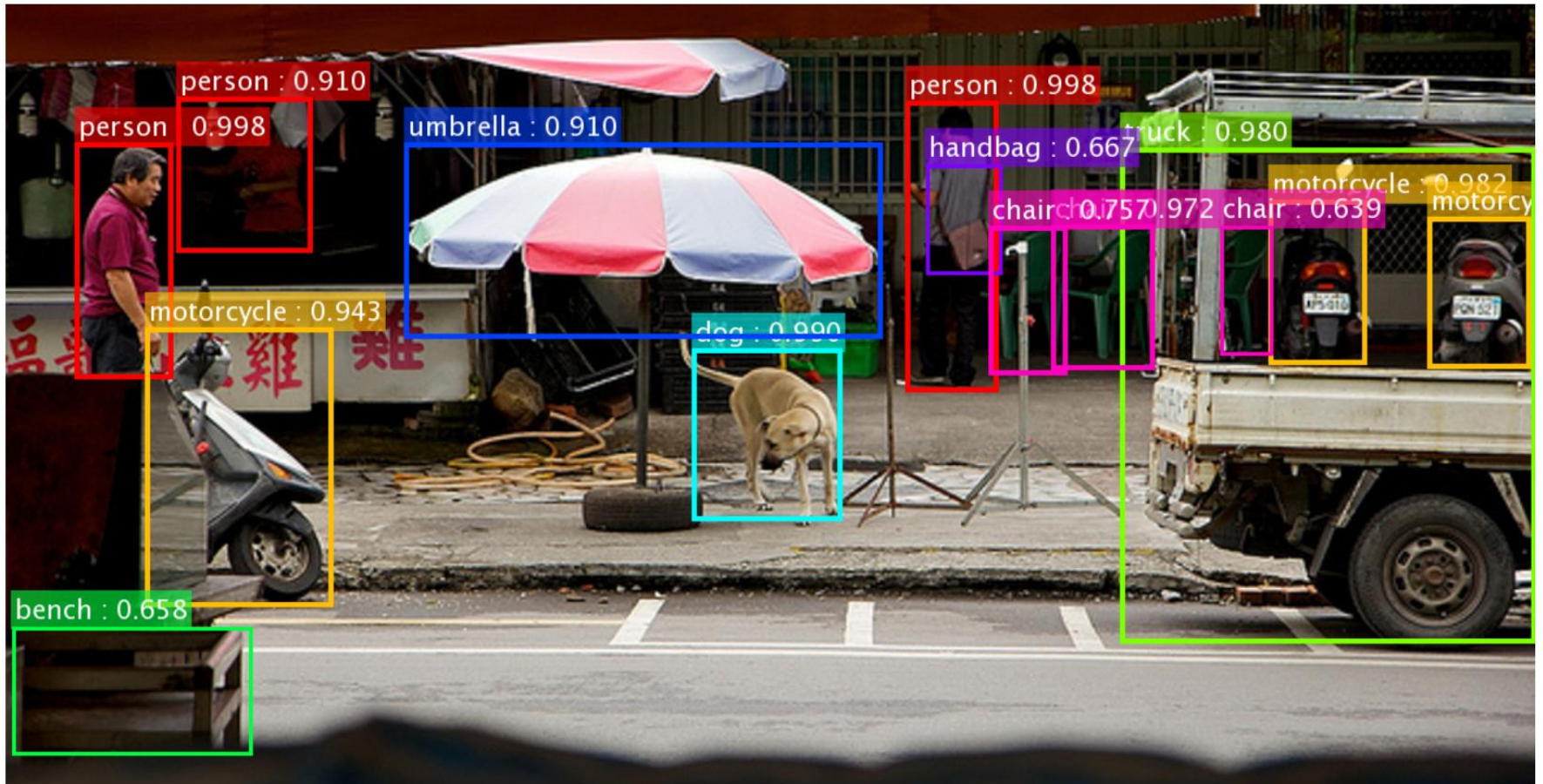
3. Distributed Deep Learning

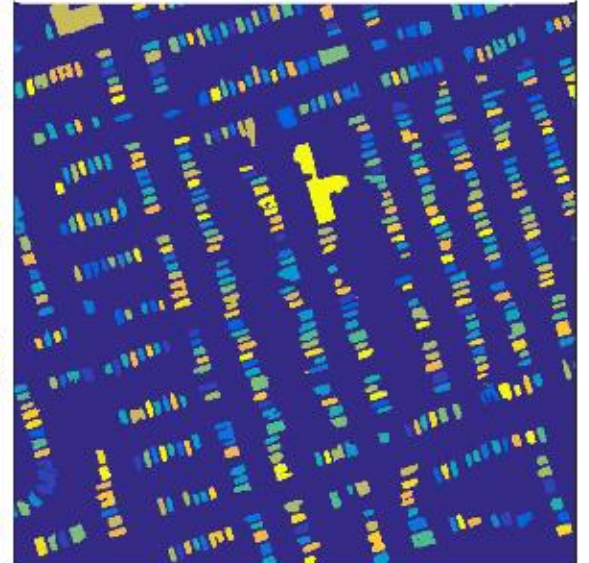
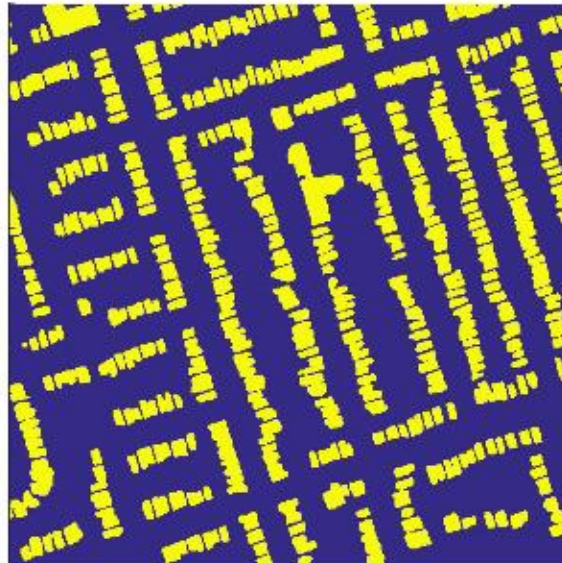
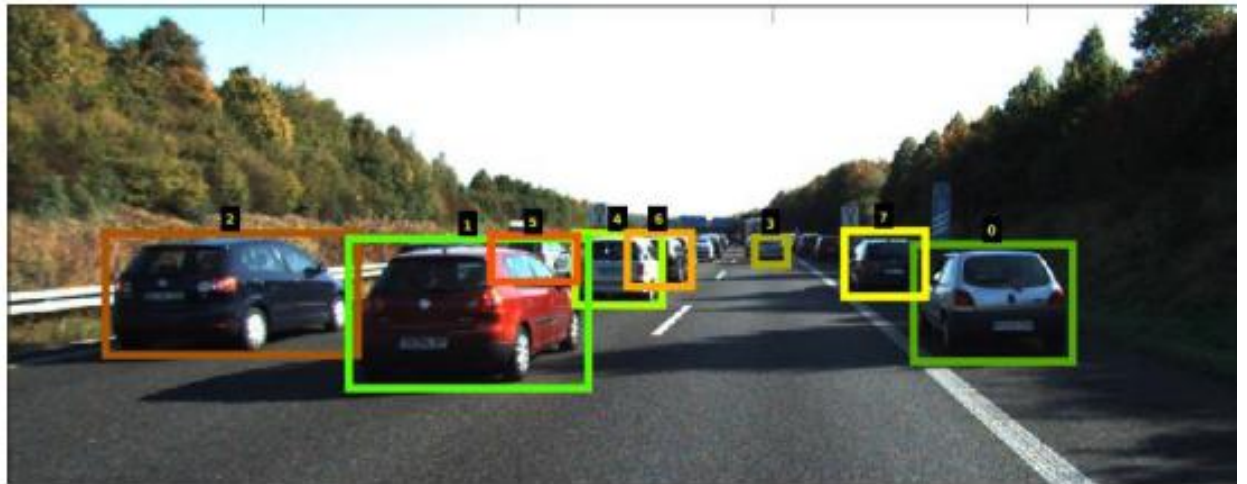
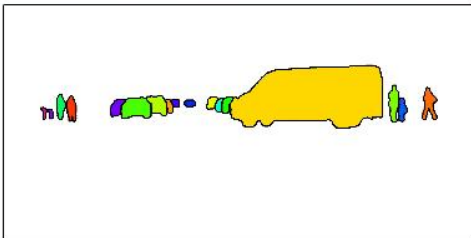
4. Conclusions

• Problem Statement

- ML demands access to very large sets of training data
- Significant problem may arise in real-world **defense** ML applications
- To overcome problems of small-data, novel **synthetic data generation** frameworks for massive ML training system are required







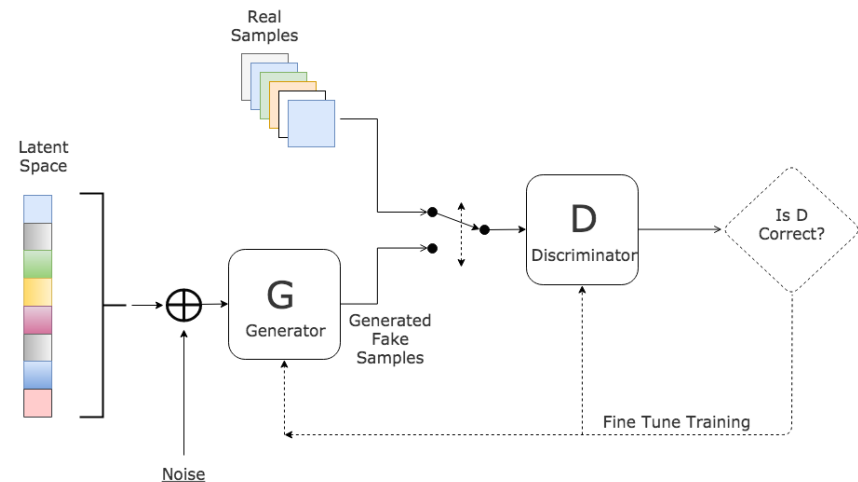
- **Objectives & Motivation**

- **Develop a massive ML training dataset generation/synthesis system**
 - 1) **Automatic Data Augmentation framework**
 - 2) **3D CAD modeling based synthetic/virtual image datasets generation framework**
 - 3) **Game Development Engine based virtual scenes/military env. creation framework**
 - 4) **Generative Models(GANs) framework for photorealistic images tuning**

• Generative Adversarial Networks (GANs)

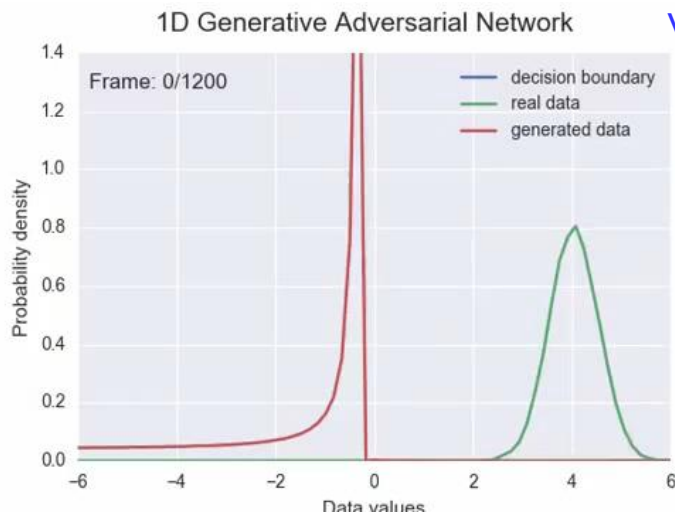
- **Data** : a dataset as samples from a true data distribution
- **Generator** : Random noise as input, and transforms it, outputting a sample of data
- **Discriminator** : Classifies sample input data as real(from true data distribution) or fake(from the generator), $x = g(z)$

Play a song on guitar
- Song (Data)
- Hands/brain (G)
- Ears/brain (D)

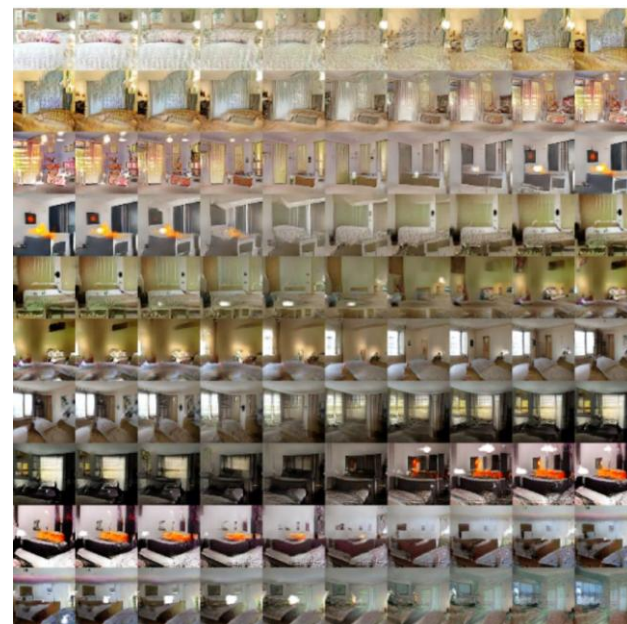


• GANs

Latent
Space
Variation



AYLIEN 2016



volcano



junco

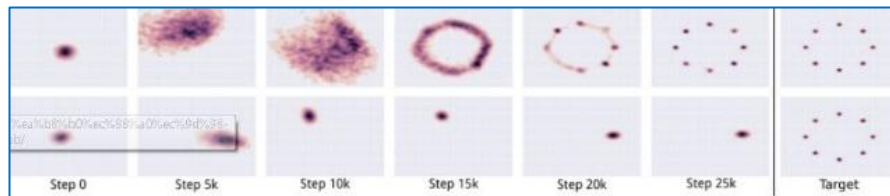
PPGN 2016 / Evolving AI Lab



Yann Le Cun 2017.3 China

- **Difficulties**

- Mode collapsing
- Oscillating during training
- No learning



CycleGAN
Ian Goodfellow
GTC 2017
(Zhu et al., 2017)

- **GANs-based Image Generation**

- **Dataset: 1,300 images of Army-tank (ILSVRC2012)**
- **Model: DCGAN**



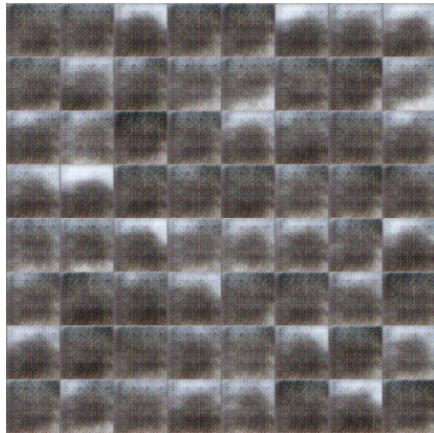
**Real army-tank training
images(ImageNet DB)**



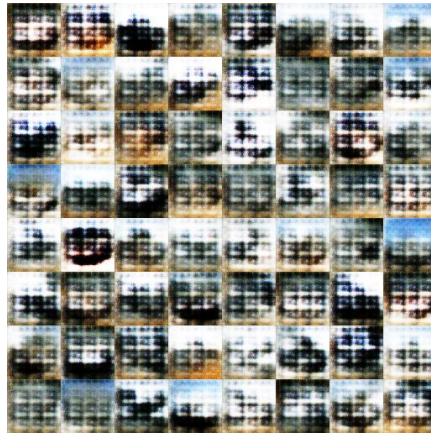
**Generated synthetic army-
tank images (Epoch=189)**

- **GANs-based Image Generation**

- **Changes of generated synthetic images along with iterations**



Epoch=5



Epoch=26



Epoch=99



Epoch=189

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- **Grade Examination**



50 EA

200 EA

10,000 EA

- **Grade Examination**



- **Employee Mgt. : absence/medical/missing/...**
- **General Management**
 - **Management/processing/copying/data missing/summary/reporting/...**
 - **Distributed Computing**
 - **Big Data Computing : Apache-Hadoop S/W**

• Sizes of data units

ZB/2yr/2배/40ZB(2020) 57*Sand

Name	Equals to	Size in bytes
Bit	1 bit	1/8
Nibble	4 bits	1/2
Byte	8 bits	1
Kilobyte	1024 bytes	1024
Megabyte	1024 kilobytes	1,048,576
Gigabyte	1024 megabytes	1,073,741,824
Terrabyte	1024 gigabytes	1,099,511,627,776
Petabyte	1024 terrabytes	1,125,899,906,842,624
Exabyte	1024 petabytes	1,152,921,504,606,846,976
Zettabyte	1024 exabytes	1,180,591,620,717,411,303,424
Yottabyte	1024 zettabytes	1,208,925,819,614,629,174,706,176

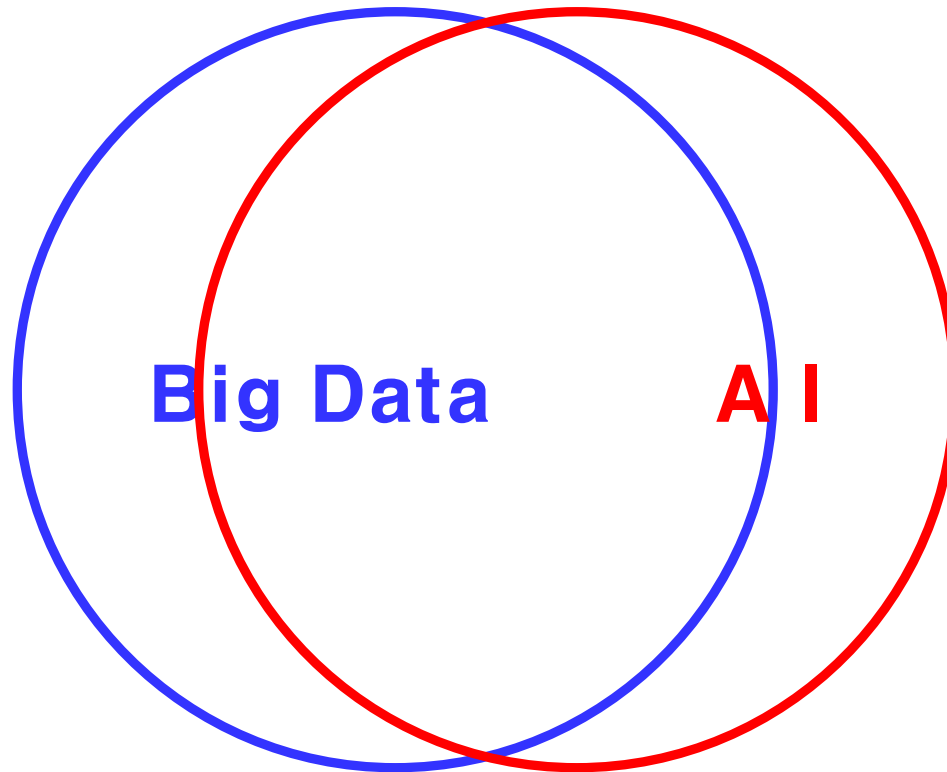
24PB Google/Day/2012

1.2ZB/2%/2013

– 1TB (@ 80 Mb/s)

- 1 disk – 3.4 hr
- 100 disks – 2 min
- 1000 disks – 12 sec

- **Big Data / AI**



- **RDB**



Library

- **Big Data DB**



Logistics

- **Big Data Component (Hadoop)**

Infomatica/Talend/InfoShere/DataStage/**Apache.Hadoop/Spark/Avro/Ambari/Chukwa/Mahout/Tez/Yarn/Zookeeper/Pig/Hive/MapReduce/Hcatalog/Impalra/Kafka/Hbase/SINGA/Stinger/MapR/Aster/Splunk/Cassandra/Mongo/Riak/Vormic/Hana/Esper/Oracle/Tibco/R/SAS/SPSS/Tableau/Spotfire/D3/Qliktech/MicroStrategy/Exadata/Greenplum/Netizza/Radian6/SMA/CloudBigdata/Amazon/Softlayer/Rackspac/Cloudant/Dynamo/Teradata cloud/ ...**



- **Apache Spark**

- 2012.06 : V.0.5.1

- ...

- 2014.05 : V.1.0.2

- 2015.09 : V.1.5.0

- 2015.10 : V.1.5.1

- 2015.11 : V.1.5.2

- 2016.01 : V.1.6.0

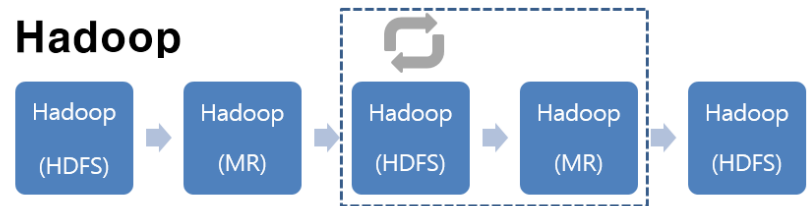
- 2016.03 : V.1.6.1

- 2016.07 : V.2.0.0

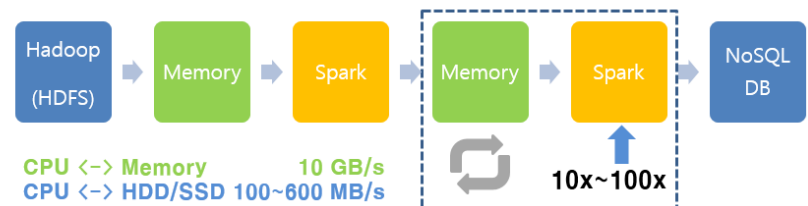
- 2017.10 : V.2.1.2



- **Hadoop**



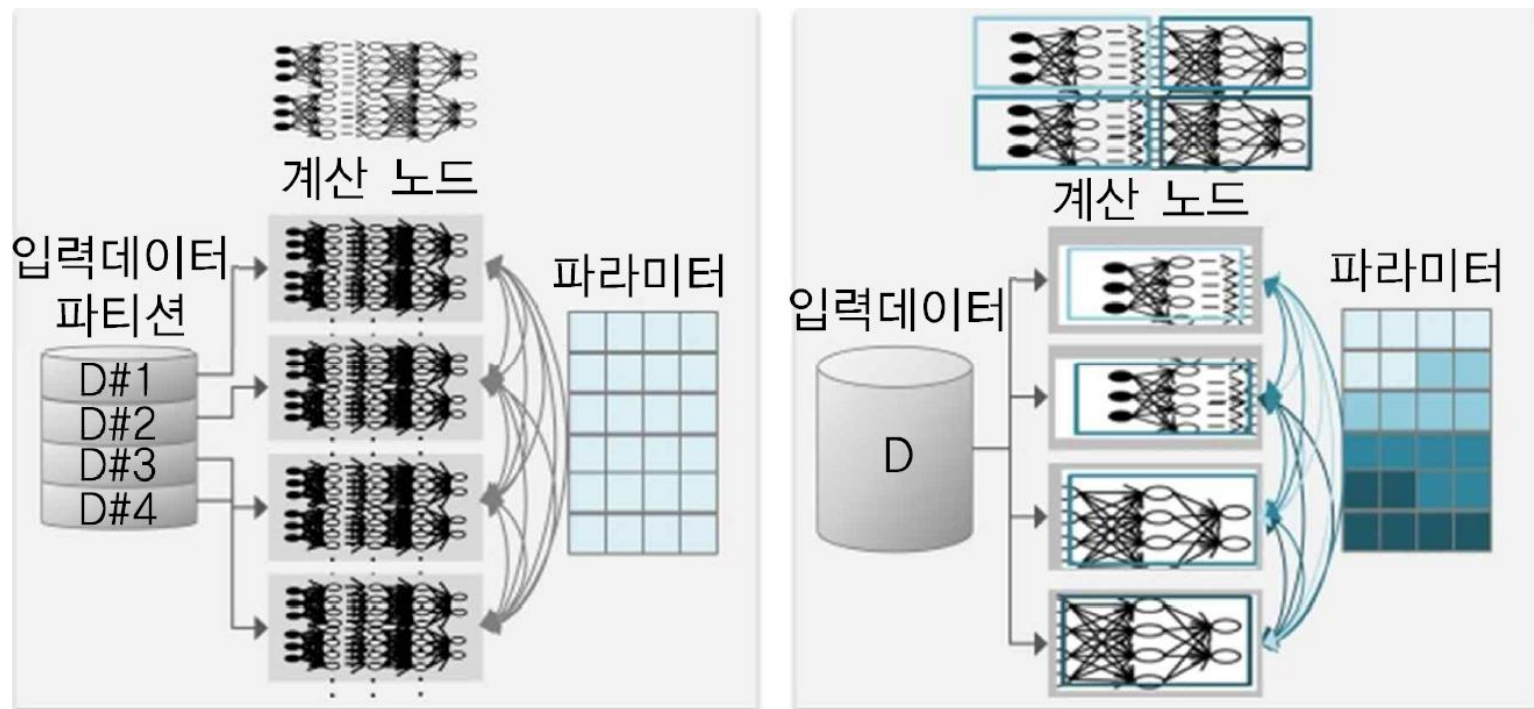
- **Spark** (from UC Berkley AMPLab, 2011~, 6년, 100명)



Google TensorFlow ('15.11)

- **Distributed Computing**
 - Data/Model parallelism
 - Full-mesh/Star topology

ETRI E/T Trends
(안신영 et al. 2016)



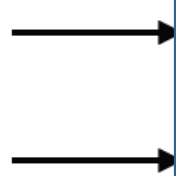


- 1,200,000 Image
- 1,000 Classes

[Image Big Data : ImageNet DB]

Image(x)

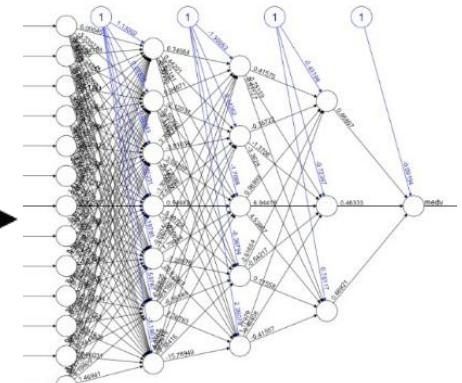
Label(y)



**Computer
(Training)**

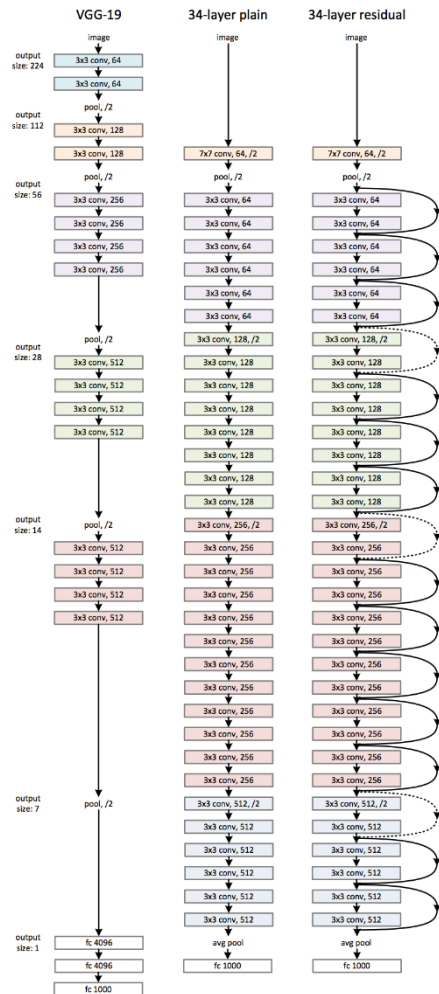


Trained w, b



NN Model

DNN Error w.r.t Hidden Layer



DNN	Layer	Error	Year	Remark
-	-	25.80%	2011	
AlexNet	8	16.40%	2012	
VGG	19	7.30%	2014	
GoogleNet	22	6.70%	2014	
ResNet	152	5.50%	2015	
	200	4.80%	2016	◎
	1,000	4.62%	2016	

Benchmark DNN

- ResNet-200 (2016)
- ImageNet (1.2M, 1,000Class)

200 Layers

Computing Time w.r.t GPU

		GPU	Petaflops (FP32)	Time	Remark
Case 1	2016	40	0.4	4.2 Days	
Case 2	–	2,000	20.0	2 hrs	
AIST-AIRC (J)	2015	8	0.1	16.8 Days	0.7 PF * 90(P)
	2018	6,500	65~100	37 min	

Tsubame 3(J)
HPE 540 Node
Tesla P100 : 2,160

- HPC with 8 GPU → 3 Weeks (ResNet 200 Layer)
- ImageNet 1.2M RGB Images with 1,000 classes

200,000 USD for 8-GPU HPC

- **BD/DL System : Open Source Platform**

Architecture	• Component
Applications	• Vision / Speech / Behavior / ...
DL Framework	• Caffe, Torch, Theano, DL4J, CNTK, Pylearn2, TensorFlow, KERAS, ...
Libraries(SDK)	• cuDNN, NCCL, IMSL Lib., GPU VSIPL, cuBLAS, cuSPARSE, cuFFT
System Software	• Apache Hadoop, Spark, Ambari, HBase, Yarn, Drivers, CUDA, ...
Hardware	• Clustered Computer (Linux) • X86, GP-GPU

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 - **Synthetic/virtual data generation**

- **Computing power**

- ML/DL based on HPC & GPU
 - Big data hub & data analytics
 - **Distributed deep learning**

