

인공지능, 퓨어스토리지를 만나다.

2018.01.09

퓨어스토리지 홍준혁



Artificial

“Artificial is Made or produced by human beings rather than occurring naturally, especially as a copy of something natural.”

Intelligence

“Intelligence is the ability to think, reason, and understand instead of doing things automatically or by instinct.”

A Proposal for the Dartmouth Summer Research Project on Artificial Intelligence

August 31, 1955

*John McCarthy, Marvin L. Minsky,
Nathaniel Rochester,
and Claude E. Shannon*

Google Trend (2012 ~ 현재)

인공지능



2012

2017



언어이해

감정표현

음성인식

이미지 인식

언어표현

인공지능 로봇 (Human-Like) 프로세스

인식 (Perception)

이해 (Understanding)

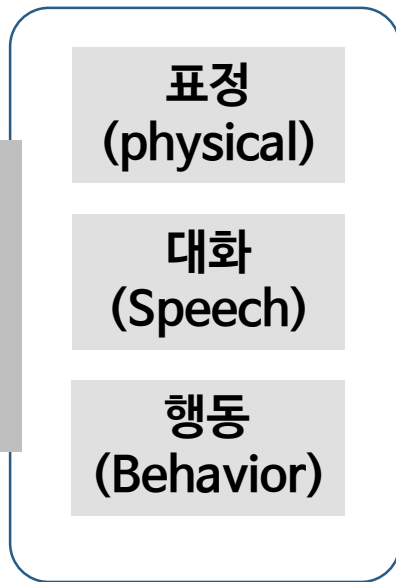
행동 (Action)



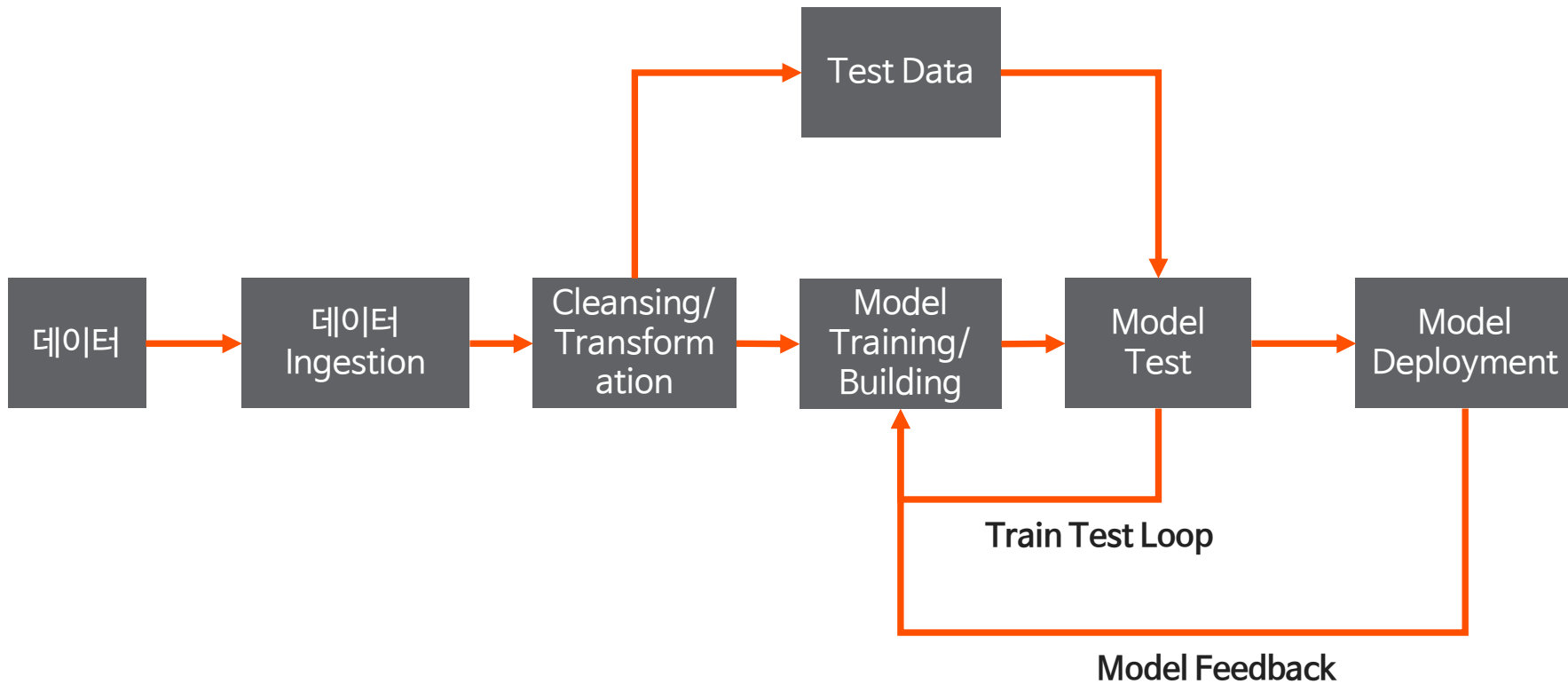
통합/합성 (Synthesis)



결과 (Result)



Learning 과정



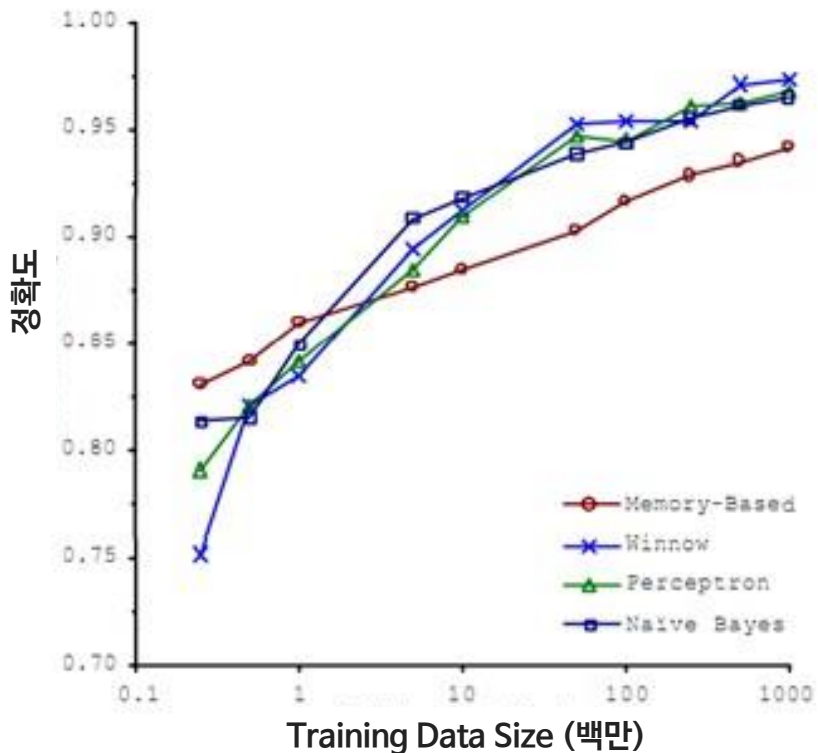
Engineering vs. Data

다양한 방식/방법 Learning Algorithm

- Memory Based
- Winnow
- Perception
- SVM
- ...

“ It's not who has the best algorithm that wins. It's who has the most data. ”

Andrew Ng

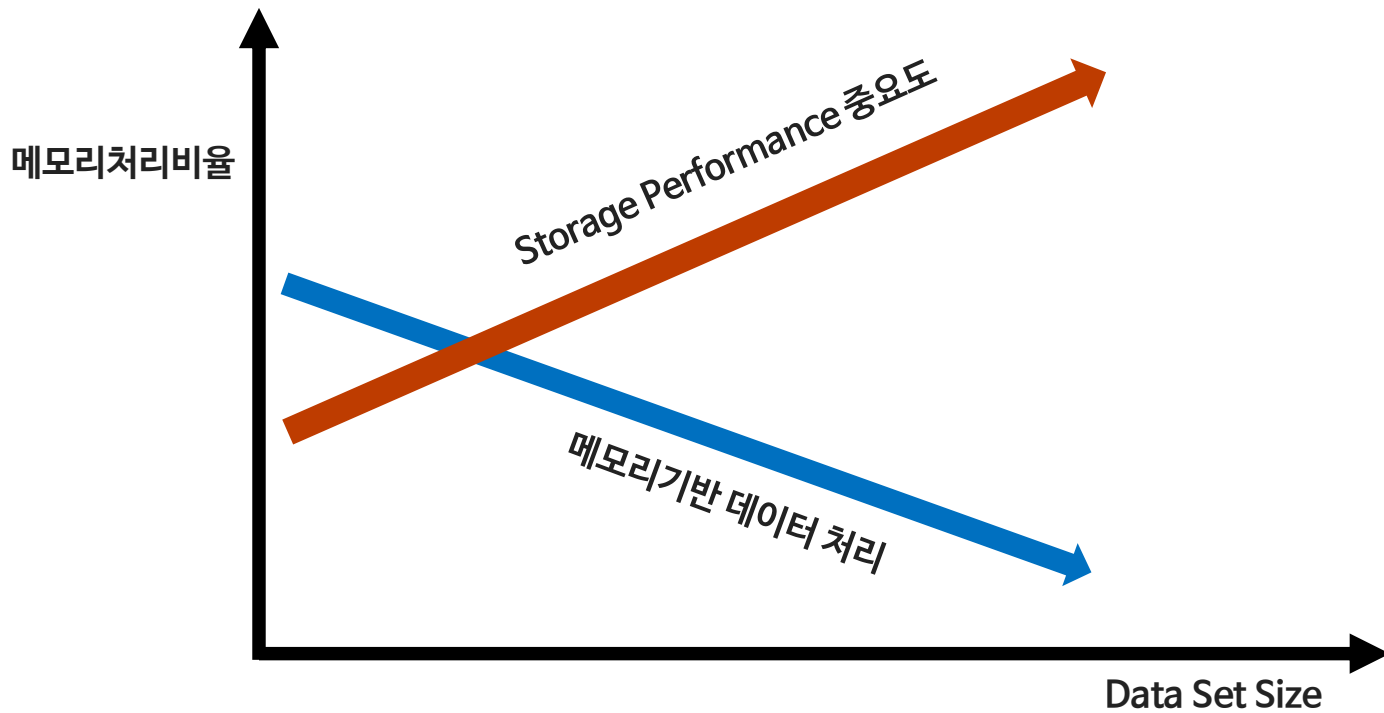




“We don’t have better algorithms. We just have more data.”

PETER NORVIG
Google Research Director

데이터 활용증가에 따라 데이터플랫폼 중요성

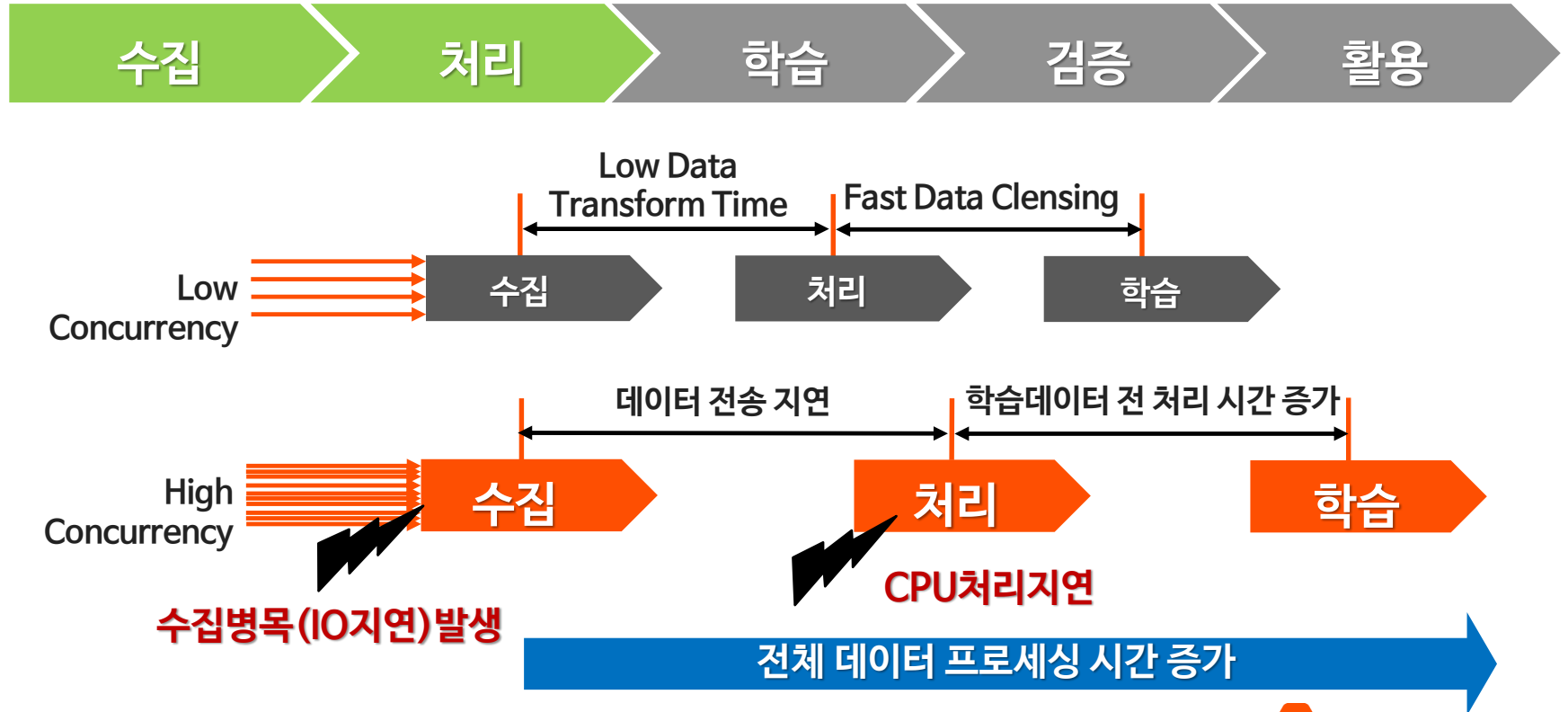


인공지능 데이터 프로세싱 과정별 특징



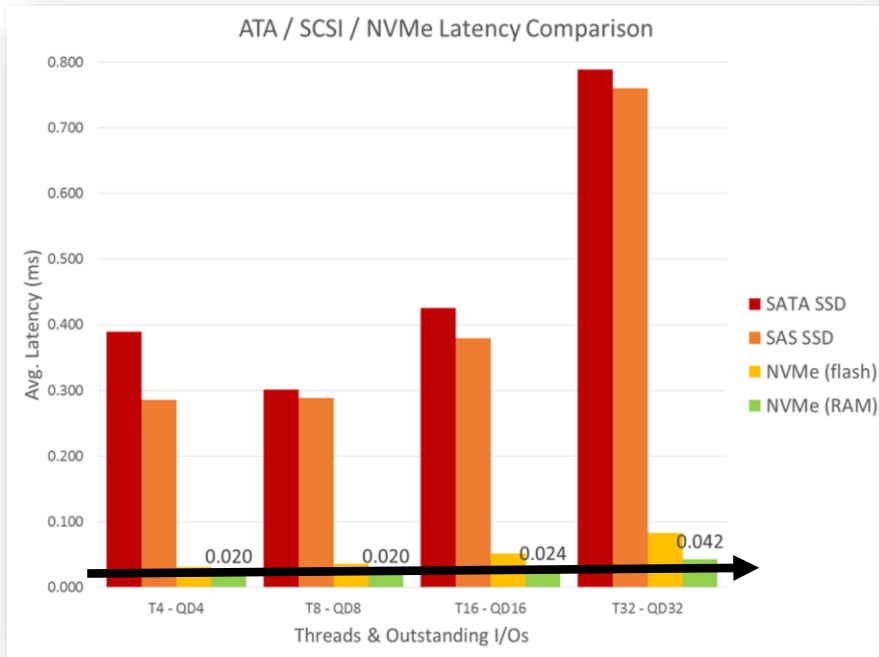
데이터 접근형태	sequential	sequential or random	random	random
데이터 접근방법	write	read & write	read	read
데이터 요청/처리 사이즈	metadata is small data is small to large	small to large	small to large	small to large
프로세스 활용	IO Bound	CPU	GPU	CPU
동시 사용 프로세스 수	depends on # of sources	high	High	Depends on # of target

수집,처리 단계의 이슈

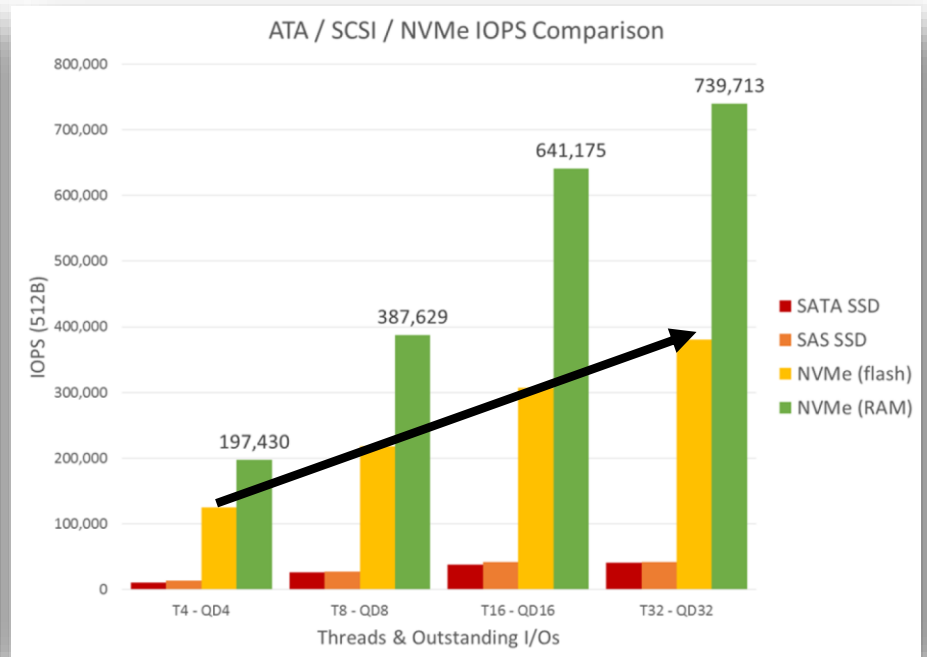


수집,처리 단계 개선 - 새로운 데이터 관리 기술

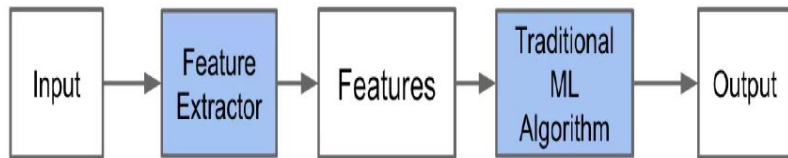
안정적인 응답속도 보장(Stability)



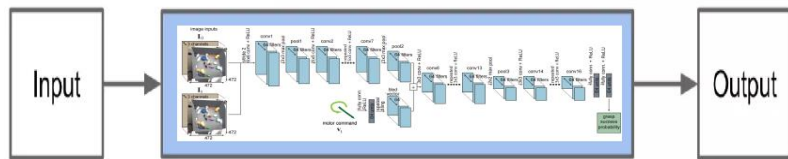
데이터 처리능력 확보(Performance)



학습, 검증 단계 이슈

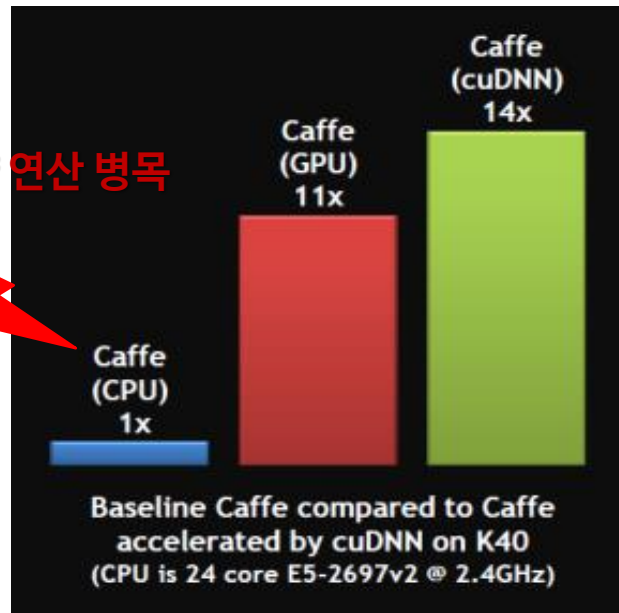


Traditional Machine Learning Flow



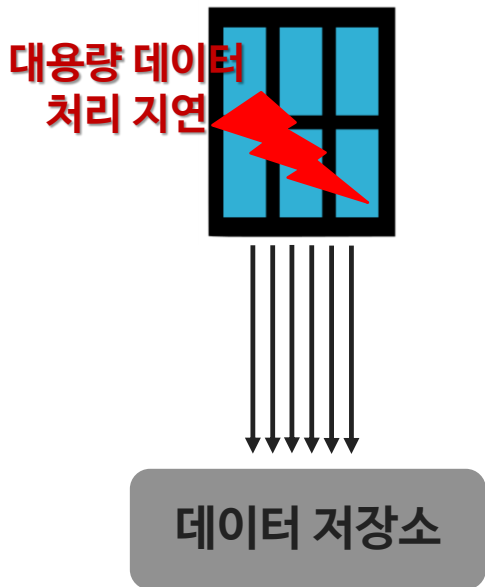
Deep Learning Flow

CPU 대응량 연산 병목

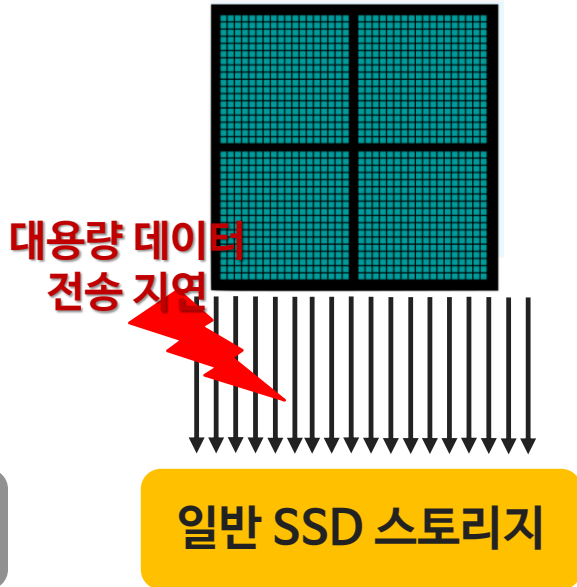


빠른 학습, 검증을 위한 GPU처리 능력 충족

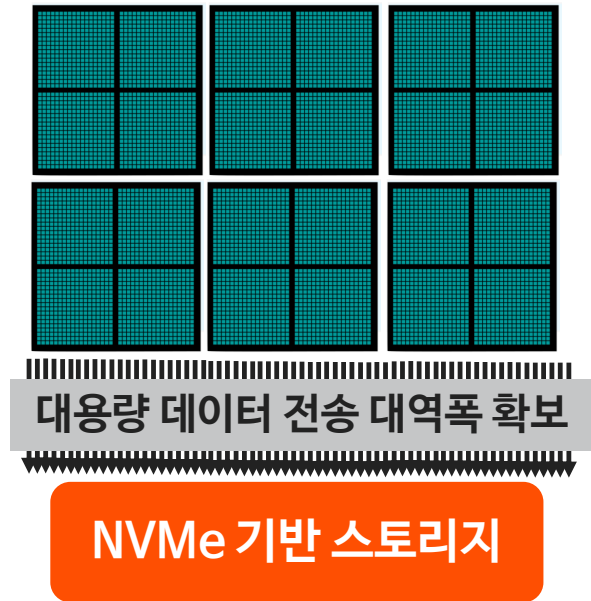
Multi-Core CPU



Thousands-Core GPU



Millions-Core GPU



인공지능을 위한 차세대 데이터플랫폼은 BOTH A TRUCK AND A RACE CAR



CAPACITY, THROUGHPUT,
SEQUENTIAL ACCESS

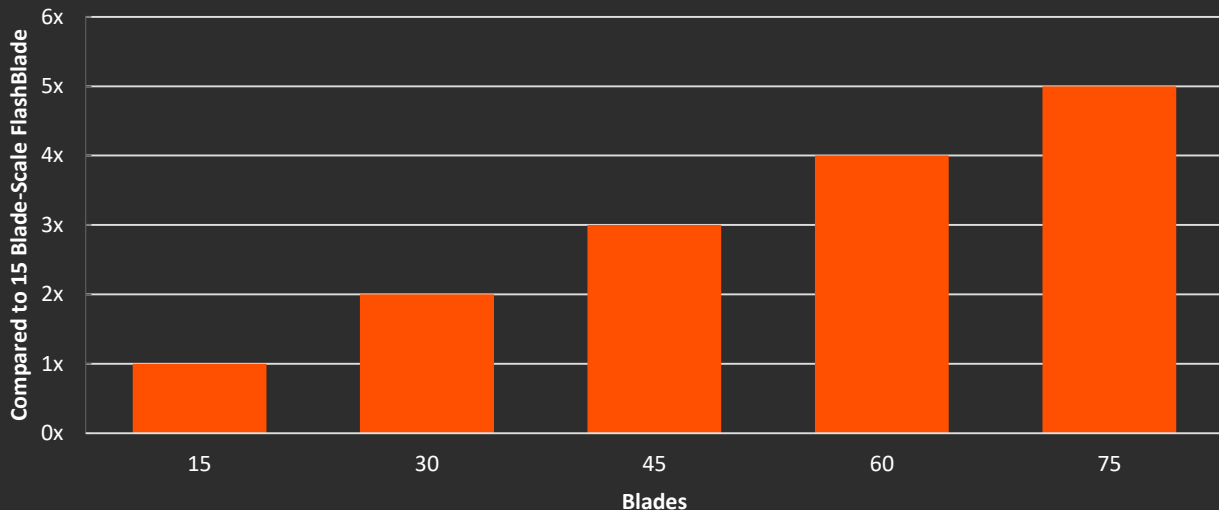


CONCURRENCY, LATENCY,
RANDOM ACCESS

FLASHBLADE

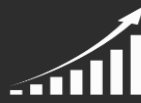
INDUSTRY'S FIRST FLASH PURPOSE-BUILT FOR MODERN DATA ANALYTICS

10PB
75GB/sec
7.5M/IOPS



FAST

Elastic Performance Up to 75 GB/s
Always-Fast, Small to Large Files
Massively Parallel from SW to Flash



BIG

10's of Thousands of Clients
10's of Billions of Objects & Files
8 Petabytes with Single IP

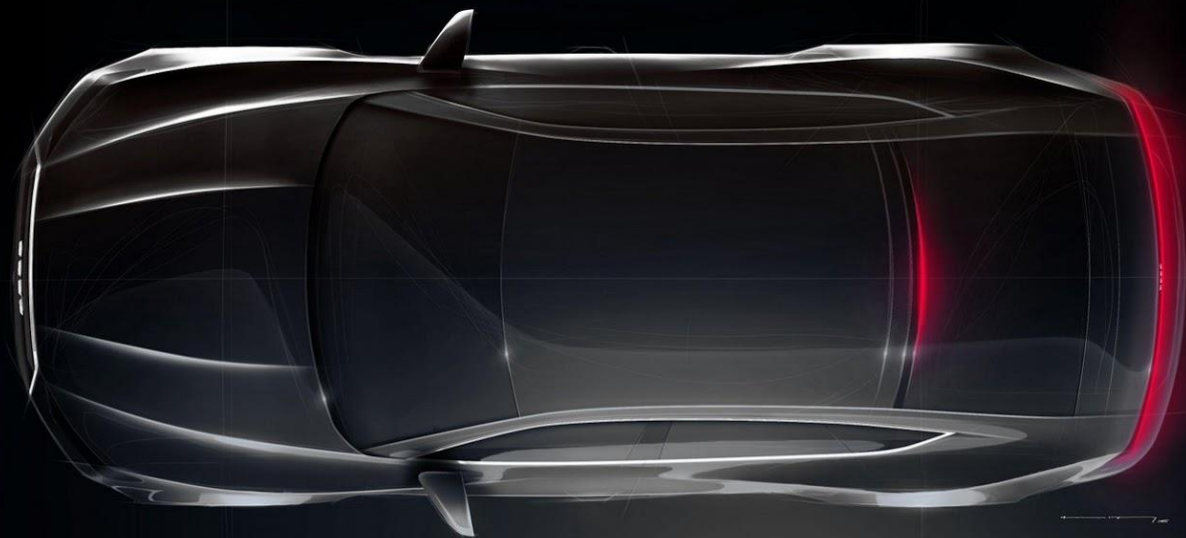


SIMPLE

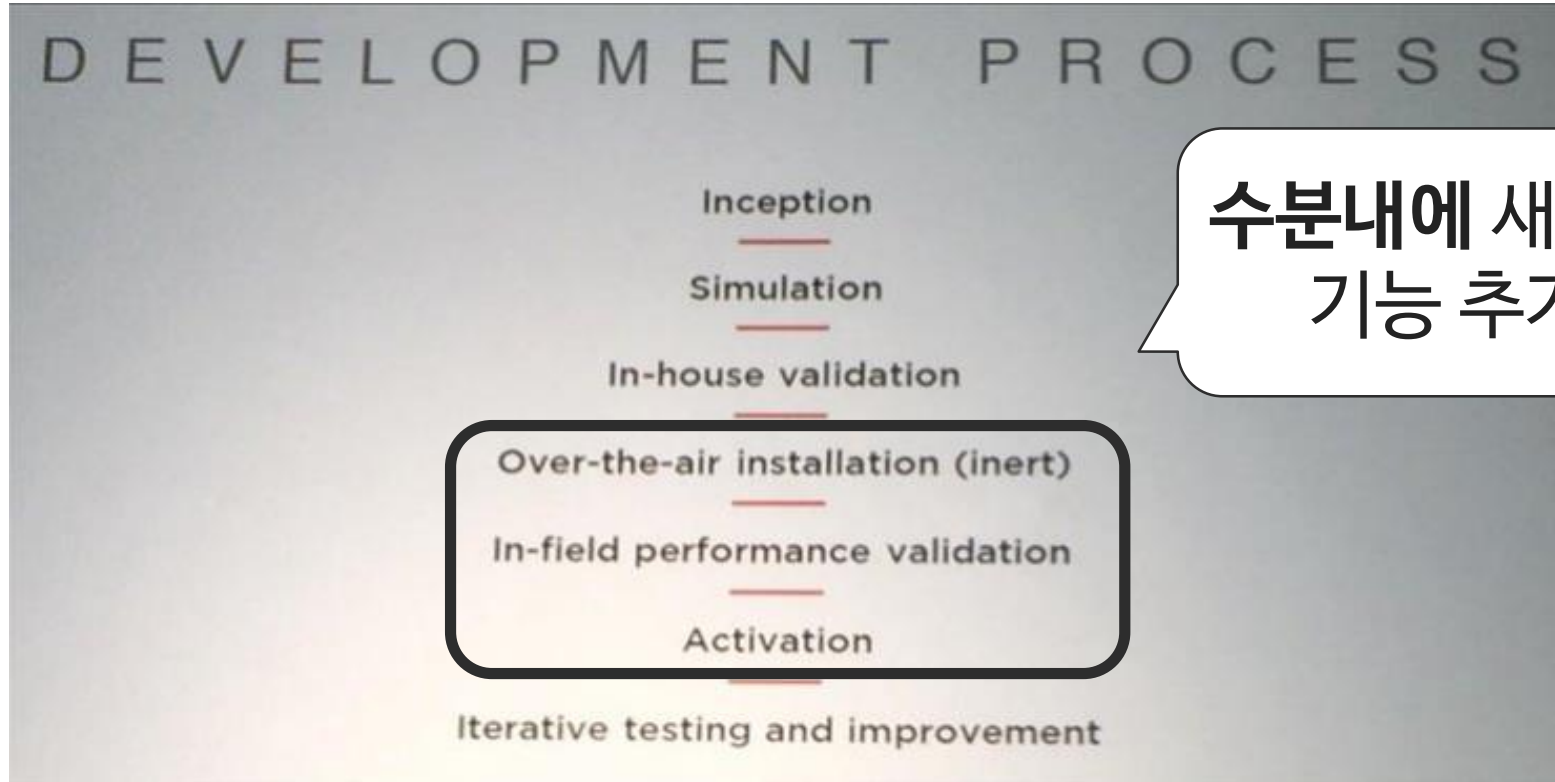
Evergreen
No Manual Tuning
Just Add Blades for Performance

퓨어스토리지를 활용한 AI 환경 구축 사례

기존 신차개발/출시 5~6 Year



인공지능기반 자율주행 자동차 신기술 적용

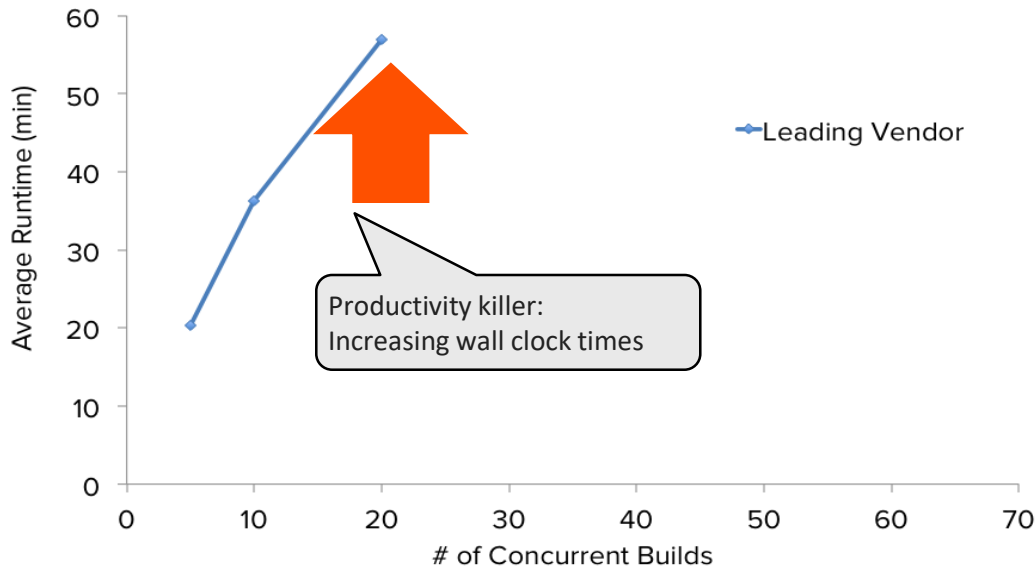


기능 개선화를 위한 개발 속도 요구

CAR Firmware Build

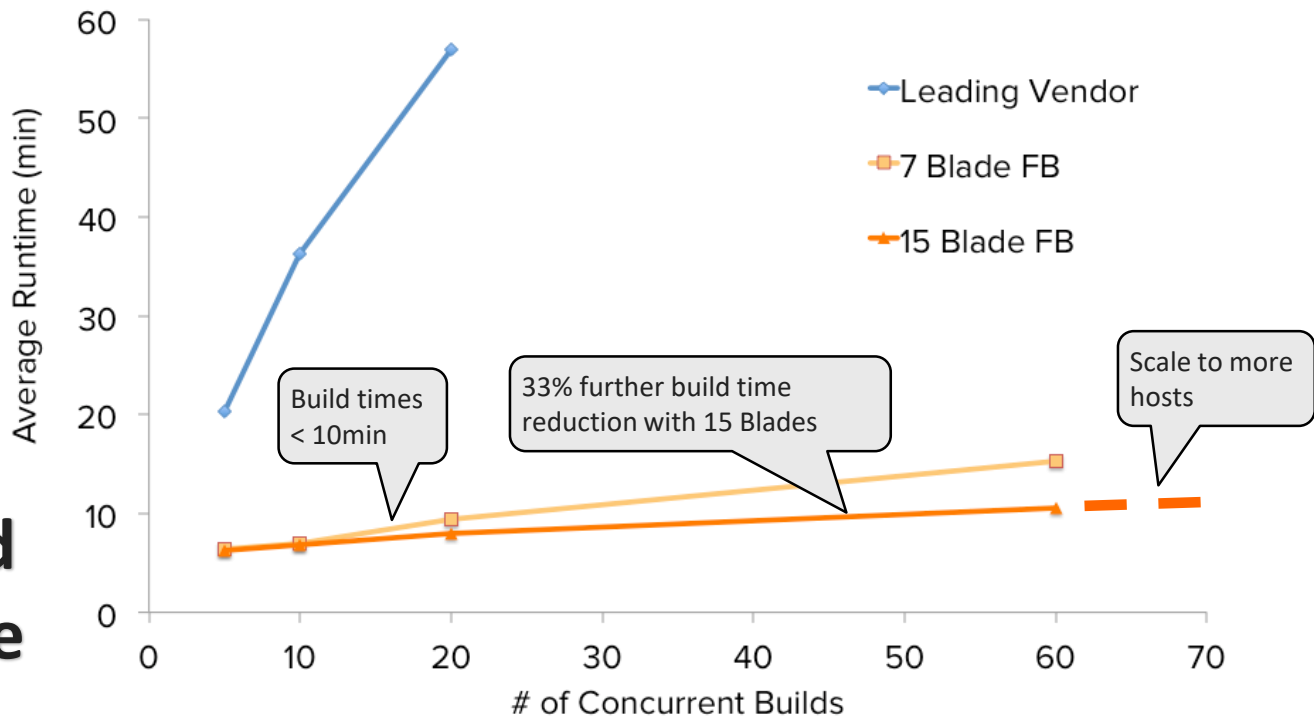
- Hundreds of SW engineers writing code
- Individual modules must be compiled together to form a full image
- Traditional Storage(NAS) systems are challenged

Concurrent Software Builds (Car Firmware)



퓨어스토리지를 활용한 개발 속도 향상

Concurrent Software Builds (Car Firmware)



80GB/sec Read
55GB/sec Write

THE COMING FLOOD OF DATA IN AUTONOMOUS VEHICLES

RADAR
~10-100 KB
PER SECOND

SONAR
~10-100 KB
PER SECOND

GPS
~50KB
PER SECOND

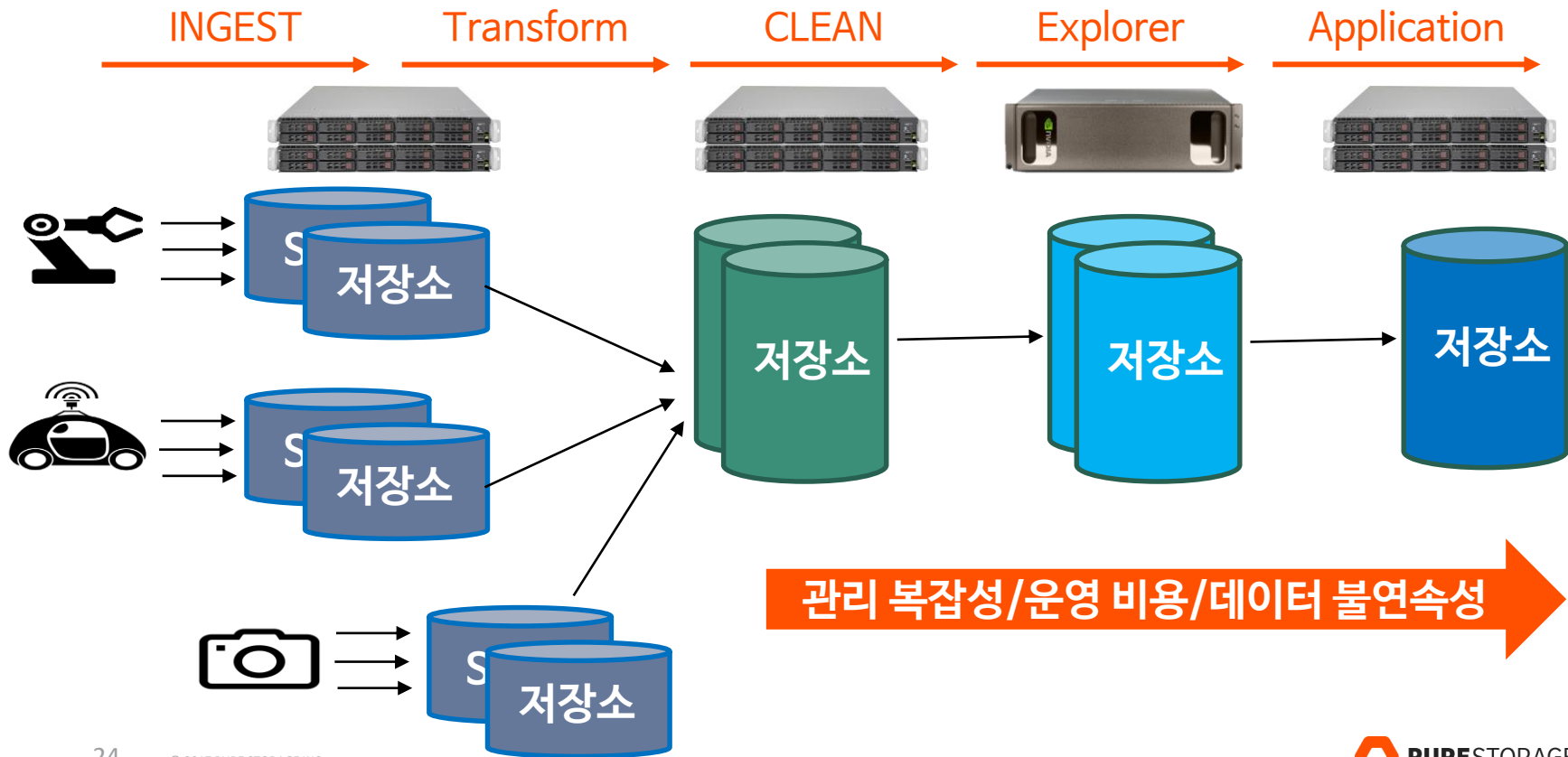
CAMERAS
~20-40 MB
PER SECOND

LIDAR
~10-70 MB
PER SECOND

AUTONOMOUS VEHICLES
4,000 GB
PER DAY... EACH DAY



다양한 데이터, 데이터 사일로 (SILO)



전문가들의 조언!!!

BUILDING A DATA PIPELINE IS CRITICAL

“Building and managing data pipelines is typically one of the most costly pieces of a complete machine learning solution.

”

Jeremy Hermann & Mike Del Balso

Uber Machine Learning Platform

<https://eng.uber.com/michelangelo/>

“If your boss asks you, tell them that I said [to] build a unified data warehouse.

”

Andrew Ng

Former head of Baidu AI/Google Brain

Nuts and Bolts of Applying Deep Learning

OLD WAY



12 Hours per Car

단일 데이터파이프라인



2.5 Hours per Car

TensorFlow-Base Product 플랫폼

Integrated Frontend for Job Management, Monitoring, Debugging, Data/Model/Evaluation Visualization

Shared Configuration Framework and Job Orchestration

Tuner

Data Ingestion

Data Analysis

Data Transformation

Data Validation

Trainer

Model Evaluation and Validation

Serving

Logging

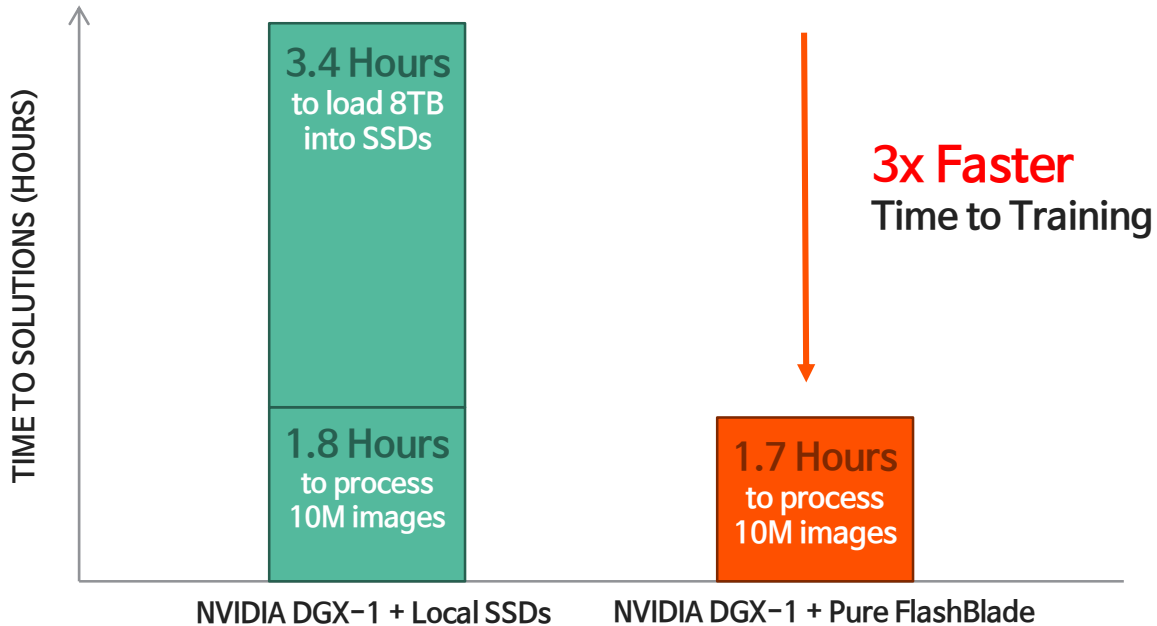
Shared Utilities for Garbage Collection, Data Access Controls

단일 데이터 파이프라인

참조 - TFX: A TensorFlow-Based Production-Scale Machine Learning Platform

데이터 파이프라인 단순화

TENSORFLOW TRAINING BENCHMARK WITH RESNET-50





Make it real.

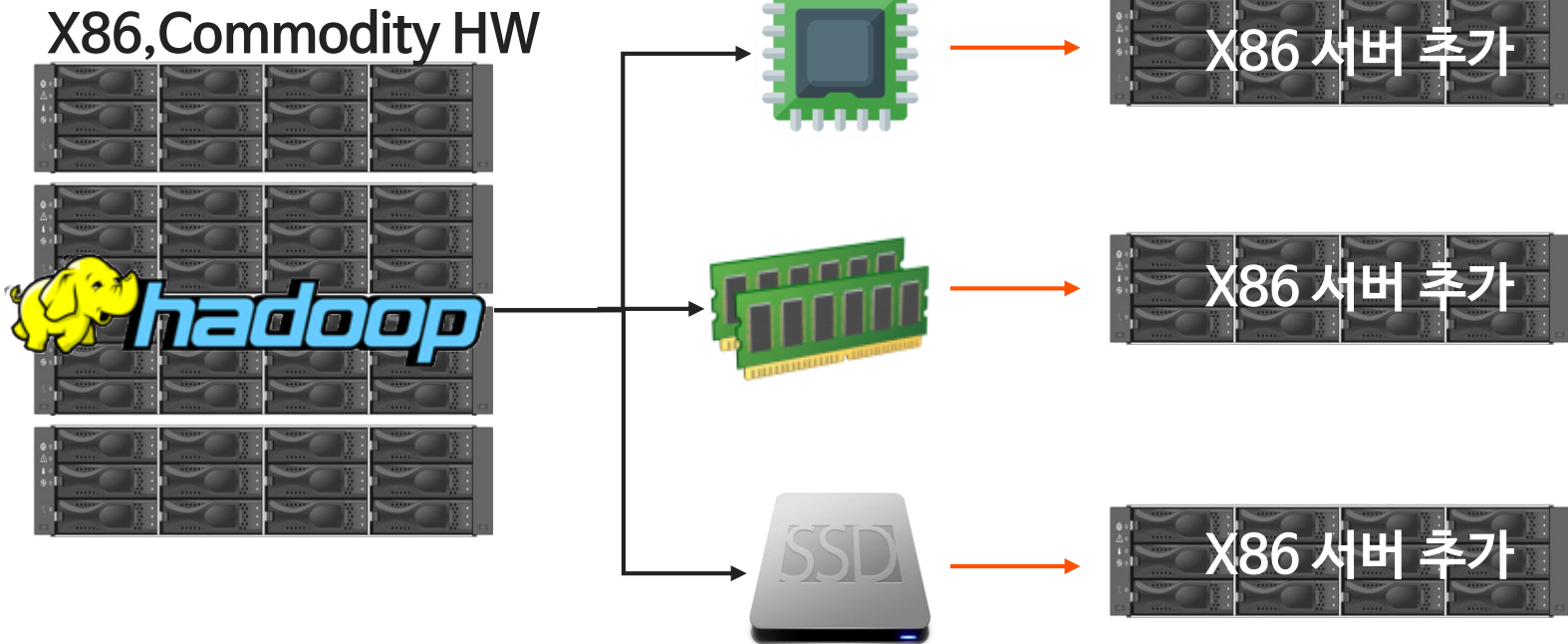


**MAKING AUTONOMOUS CARS
POSSIBLE BY 2021**

Zenuity, a joint venture of Volvo and Autoliv, chose NVIDIA DGX-1 and Pure FlashBlade systems for their deep learning infrastructure.

빅데이터플랫폼, 리소스 증설 한계 발생

Tightly-Coupled 아키텍처



빅데이터플랫폼, 클러스터 관리 복잡



HBASE

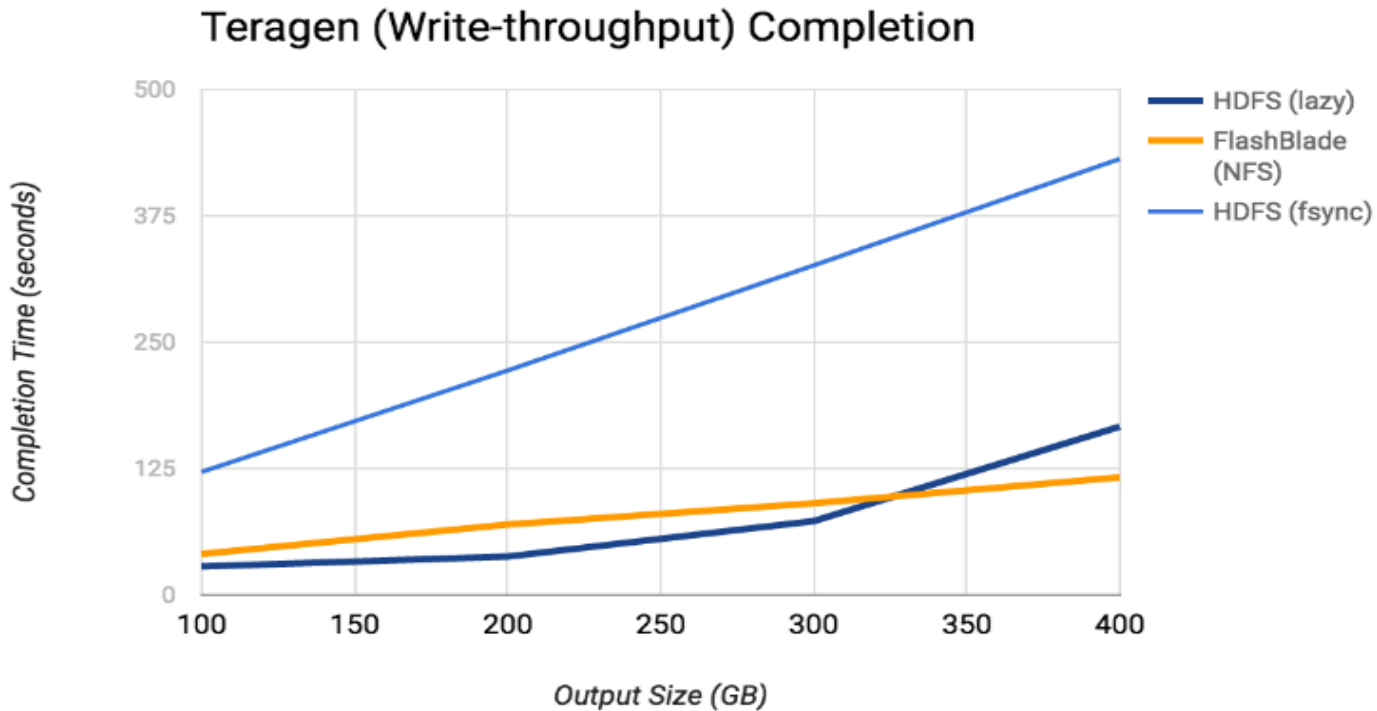


단일 장애가 다중 장애로 확산

결론, 핵심역량에 집중



FLASHBLADE, HDFS 성능 및 안정성 향상



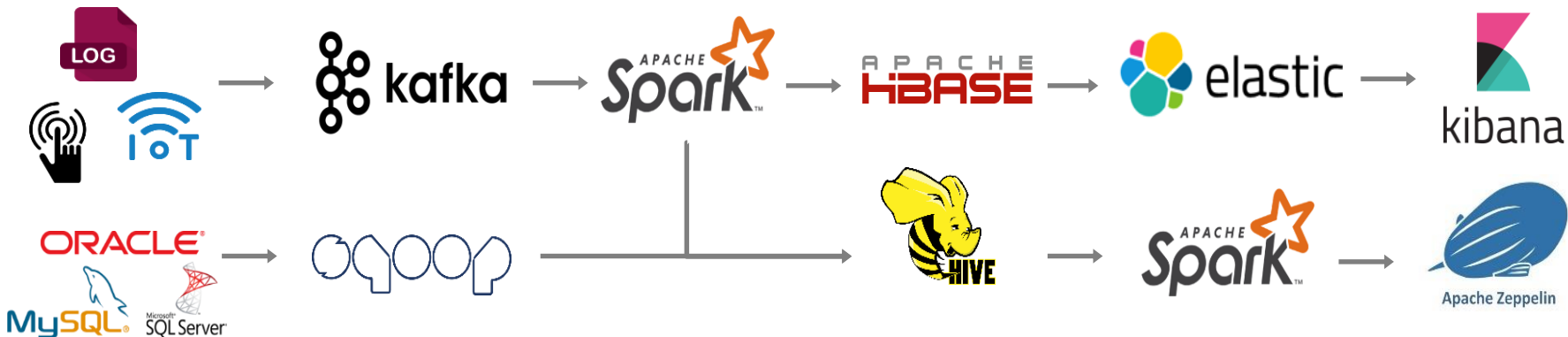
10X FASTER INVESTMENT DECISIONS WITH PURE FLASHBLADE



“Our quants want to test a model, get the results, and then test another one – all day long.”

Gary Collier, co-CTO, Man AHL

람다 아키텍처 (실시간+배치) 분석 플랫폼



공유 파일 시스템 (NFS) 기반 고가용성, 단일 관리 HDFS구성



EDA, HPC기반 고성능 시스템 필요

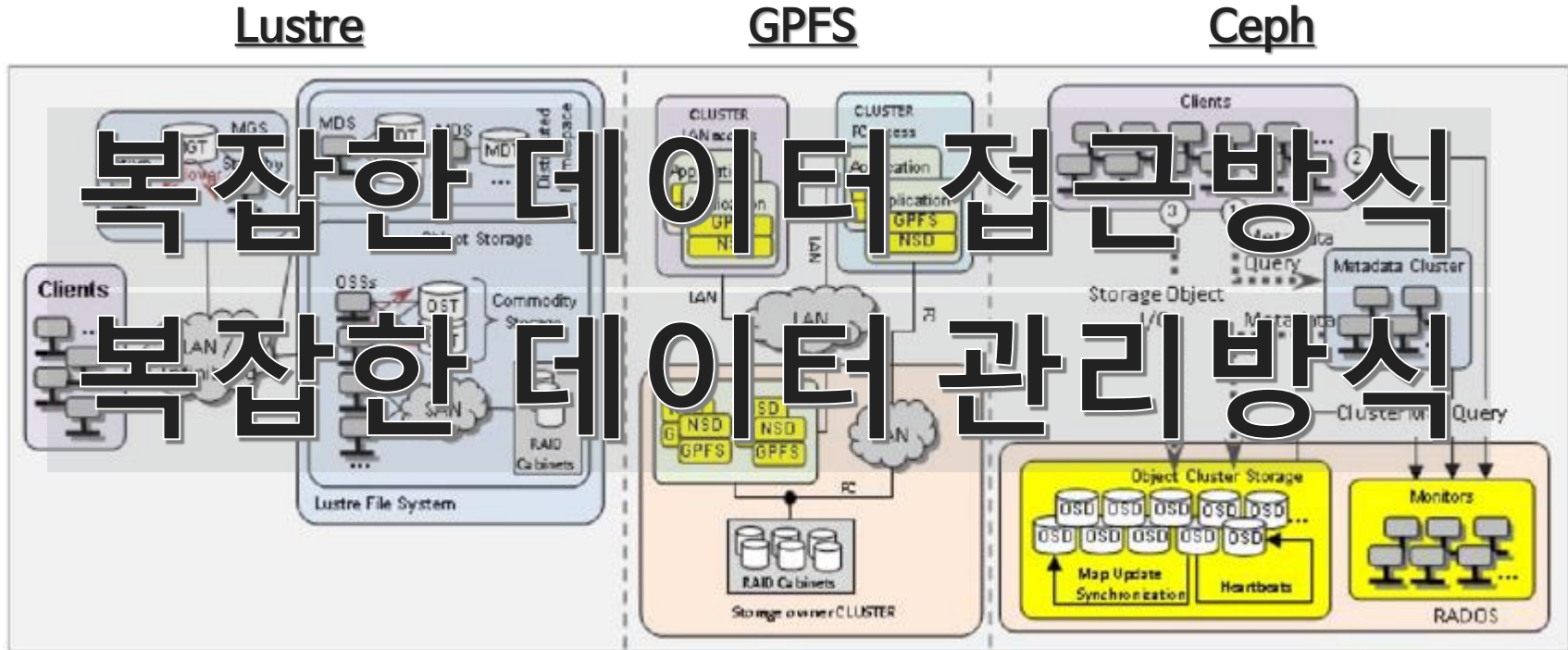
- ✓ 수 백대 혹은 수 천대 서버
- ✓ 분산 컴퓨팅
- ✓ 대용량 네트워크 대역폭
- ✓ 동시 접속 사용자
- ✓ 목적에 따른 다양한 툴

High Throughput Computing

High Performance Computing

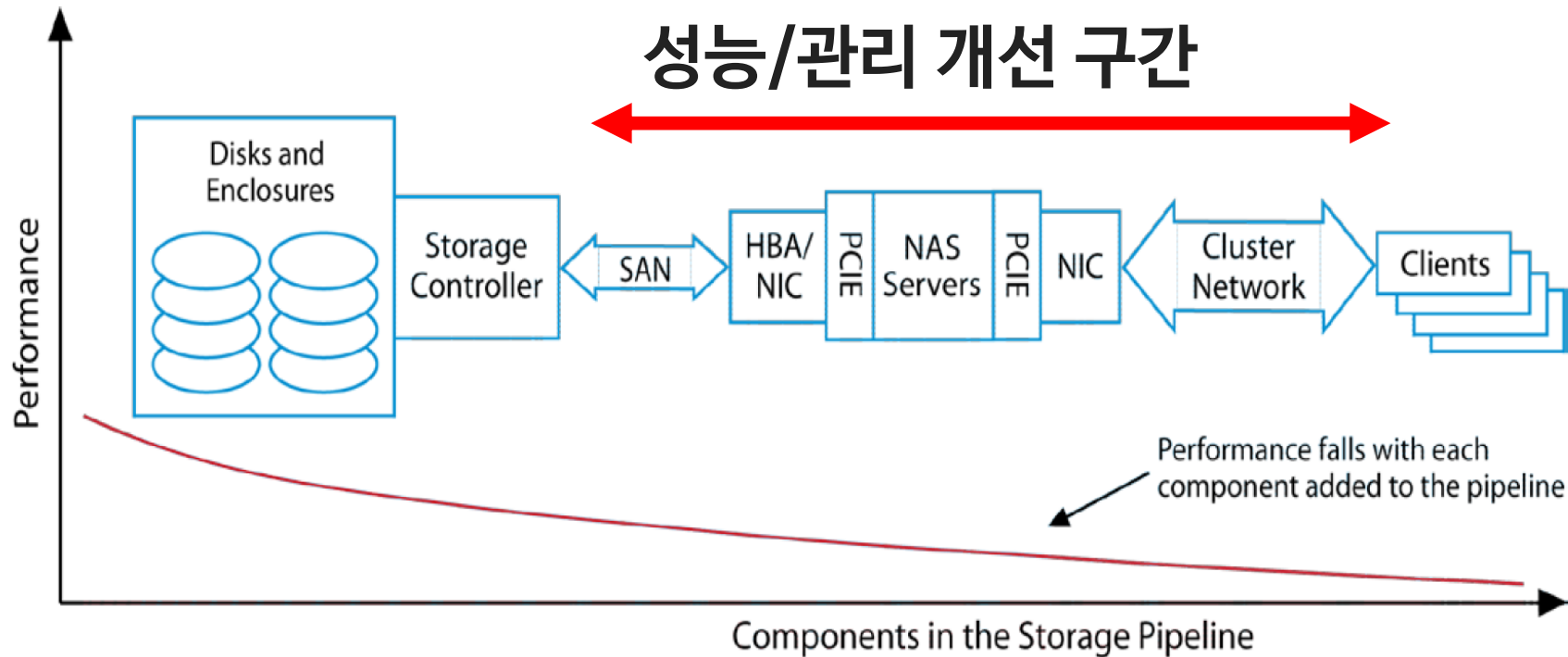
High Concurrency Computing

복잡한 데이터 아키텍처



https://www.researchgate.net/figure/278847153_fig1_Fig-6-a-Lustre-Architecture-b-GPFS-Architecture-c-Ceph-Architecture

성능, 관리를 위한 데이터 접근 단순화 필요



고성능 컴퓨팅 스토리지 3가지 요구 조건

1 Linear Scalability보장

- The ability to scale out in a clustered architecture to cater for increased I/O and capacity requirements.

2 Object 기반 데이터 관리

- The use of [object storage](#) technology, which organises files as objects in a flat directory rather than a traditional tree hierarchy. This boosts performance for very large collections of files because the storage system does not have to navigate a tree structure to locate a file but instead consults an index, often held entirely in memory.

3 IO처리과정 단순화

- Use of [parallel file systems](#) across multiple nodes and in some cases driven by custom silicon, which allows more rapid access to large files than when using a single node, which can become a bottleneck.

WORLD'S FASTEST RACE CARS

DESIGNED WITH PureStorage



[3.2×10^9 bp x 2 bit]



- * <http://sandwalk.blogspot.kr/2011/03/how-big-is-human-genome.html>
- * <https://www.technologyreview.com/s/427720/bases-to-bytes/>

다양한 형태의 유전자 분석 데이터를 저장

```
AGCCCCTCAGGAGTCC
GGCCACATGGAAACTC
CTCATTCCGGAGGTCA
GTCAGATTACCCTGG
CTCACCTTGGCGTCGC
GTCCGGCGGCAAAC
AGAACAC
```

```
@SEQ_ID
GATTTGGGGTTCAAAG
CAGTATCGATCAAA
+SEQ_ID(Optional)
!*(((****))%%%+)(
%%%%).1**
```

```
r001 99 chr1 7 30 17M
= 37 39
TTAGATAAAGGATACT
G |||||
r002 0 chrX 9 30
3S6M1P114M * 0 0
AAAAGATAAGGATA
|||||||6|BI NM:i:1
```



Raw Data

FASTQ

BWA

SAM

SAM Tools

BAM

BCF Tools

VCF

800MB

300GB

130GB

570GB

10G

고성능 유전자 데이터 처리능력 필요

FASTQ->VCF생성

Table 3: NGS Analysis Pipeline Configuration

Application	NGS sequence mapping		SNP Calling							
	bwa		samtools					bcftools		
Option	aln	sampe	view	sort	index	flagstat	merge	index	mpileup	view
Input	.fastq	.sai, fastq	.sam	.bam	.bam	.bam	.bam	.bam	.merged.bam	.bcf
Output	.sai	.sam	.bam	.bam.bai	.flagstat	.bam	.bam.bai	.bcf	.vcf	
Term	bwa aln	bwa sampe	samtools sort			samtools merge		samtools mpileup		

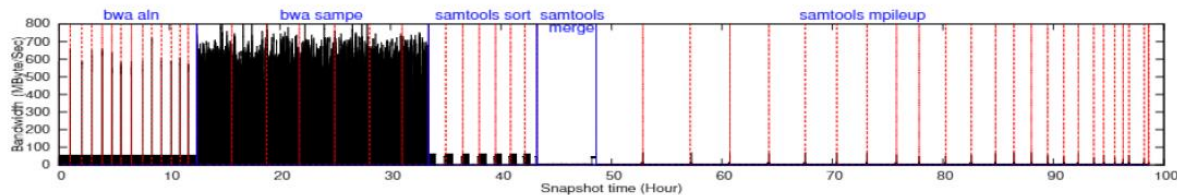
Total Data Size : 약 1100GB

Table 4: # of File Operation and Write/Delete File Size

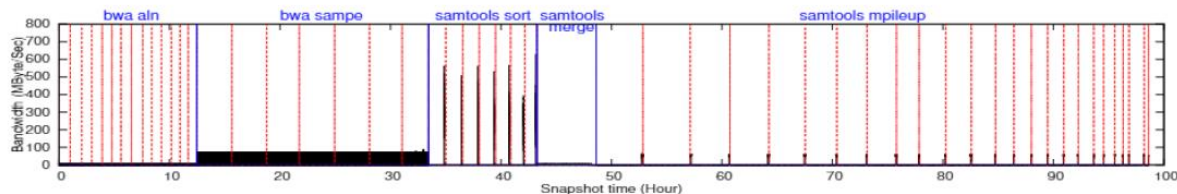
	# of Files Created/Deleted/Opnd	Write size (GByte)	Delete size (GByte)
bwa aln	14/0/2979	91.0	0
bwa sampe	7/0/1647	320.0	91.0
samtools sort	556/535/12	227.0	0
samtools merge	2/0/10	69.4	0
samtools mpileup	52/0/108	323.4	0
total	631/535/4756	1031.7	91.4

1인 분석을 위한 필요
스토리지 성능

최대 1GB/sec



(a) Read Bandwidth



(b) Write Bandwidth

NGS 다양한 워크로드 충족

PRIMARY ANALYSIS	SECONDARY ANALYSIS	INFORMATICS
Create Base Pairs And Quality Scores	Align Reads to Assemble Sequence, Map to Reference Genome and Variant Calling	Interpretation, Discovery and Reporting
SMB/CIFS 100TB+ 100GB - 200GB	NFS 10PB+ 200GB - 500GB	NFS/Objects 100PB+ GB - TB
Medium Performance Fast writes, Medium BW Scratch Single Server FASTQ files	Very High Performance High IOPS/Metadata, Low Latency, High Aggregate BW Scratch -> Enterprise Large Server farms - 100K cores BAM files, VCF files	Medium Performance Med IOPs/BW Enterprise, Hybrid/Cloud Enterprise Servers Word/PDF/proprietary

3X FASTER SPARK FOR GENOMICS

“ In our HDFS system, our cluster typically runs at 10-30% utilization of compute capacity, but at 80-85% of storage capacity... trying to **increase storage capacity built out around HDFS would have been impractical.**

We've been able to take some workloads that **we couldn't run on our existing hardware, move them to FlashBlade and run them with no problem.** While that's huge for our team, it's even more important to the lives we might impact in the future. ”

The Alconics Awards 2017

Best Innovation in AI Hardware

WINNER



The
AI Summit®
SAN FRANCISCO

