

NSML: A Machine Learning Platform That Enables You Focus on Your Models.

ML-Sys WS 2017 @ NIPS **Nako Sung**, Minkyu Kim, Hyunwoo Jo, Youngil Yang, Jinwoong Kim, Leonard Lausen, Youngkwan Kim, Gayoung Lee, Donghyun Kwak, Jung-Woo Ha, and Sunghun Kim

CLOVA AI Research (CLAIR), NAVER | LINE, Search Solution, NAVER Webtoon, HKUST



E



- A machine learning platform that enables you focus on your models
- Two options: on-premise / PaaS









The #1 programmer excuse for legitimately slacking off - 2017 version







https://www.youtube.com/watch?v=lxZyxxHOw3Y





Wasted Time



https://www.youtube.com/watch?v=lxZyxxHOw3Y







https://www.formula1.com/en/latest/features/2017/2/F1-cars-of-2017.html





Importance of Fast Machines (Multiple Servers and GPUs)



https://www.formula1.com/en/latest/features/2017/2/F1-cars-of-2017.html







https://www.sportskeeda.com/f1/what-happens-during-f1-pit-stop





ML Research Challenges: Incidental Tasks



https://www.sportskeeda.com/f1/what-happens-during-f1-pit-stop













ML Research Challenges: Resource Scheduling and Utilization





14 GPUs available but only 7 GPUs can be used in a single machine.





https://livingthing.danmackinlay.name/automl.html





ML Research Challenges: Hyperparameter Tuning



https://livingthing.danmackinlay.name/automl.html











Tensor board





Visdom



ML Research Challenges: Multiple Experiments





Tensor board





Visdom





https://www.linkedin.com/pulse/protecting-workers-who-work-alone-sandie-baillargeon





ML Research Challenges: Isolated Researchers



https://www.linkedin.com/pulse/protecting-workers-who-work-alone-sandie-baillargeon







- Slack
- Incidental Tasks
- Inefficient resource utilization
- Naive hyperparameter tuning
- Painful keeping track of multiple sessions
- Isolated researchers



Challenges



Requirements of ML Platforms

- Resource Management
 - Better computational resource management
- Data Management \bullet
 - Post datasets once and reuse them for multiple models
 - Share datasets with others
- Serverless Configuration
 - No framework / library lock-in
 - Easy and lightweight task submission



Requirements of ML Platforms

- Experiment Management and Visualization
 - Parallel runs with different jobs priorities
 - Automatic visualization and summarization of learning progress
- Leaderboard lacksquare
 - Leaderboard for each dataset to compare models and hyper parameters
- AutoML
 - Experiment performance prediction based on previously run experiments.



• Automatic hyper parameter optimization based on the performance predictions.

Limitations of Previous Solutions

- Vendor lock-in (Cloud service)
- Inefficient model experiments
- Inconsistent research environments
- Still hard to keep track of experiments



This work was done for NCSoft and was presented at Nvidia GTC Korea 2015.



0.0004 20.01









	v3 2x	of 0.5x spec	c 19		• / • ×	Best: 39e19cd68c	88.083
٦	Trainers			Testers	win% score	Q loss sampled at	-0.036195 0.003609 209688
	Q : 0.640	R: -0.022	S: 1.5% 0.009718			iter_size	2 40
	Q : 0.038	R : -0.019	5: 8.8% 0.005151			num_states replay_size target_q.sync_every unique	75 0.1 2000 0.0896063
	Q : 0.058	R : -0.021	\$: 2.1% 0.002295			uuid.39e19cd68cb7e	1
	Q : 0.033	R : -0.024	5: 0.9% 0.011015				•
	Q : 0.014	R : -0.020	\$: 4.6% 0.011176				
	Q :-0.014	R: -0.021	5: 1.4% 0.015642				





This work was done for NCSoft and was presented at Nvidia GTC Korea 2015.

My Previous Work in Early 2015

Solution	E (G	JE D		BU BU SU
Overview Settings	Status					## Status 2 ## ## Status 3 ##
global						Perfs(4, AIGhTick) consume sec : 0.0
					1 - 7	Team1(RemoteQ.1] 18 5 Team2(Evalro T1 MinPerci 78.36X (58-MA:78.26X)
Title	test					## Events ##
Algorithm	NEQ				•	
Connect UDI	conice-iteamoday	-2				
Connect UKL	service.//game/sor	15				
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Parent						
default						Status
				1		History Trainers Testers
batchsize	LDG •	16	64	16	-	39+19cd/i8do7ea1fcc5/C2a9235#adc09e0cda6e6
gamma	LINEAR •	0.94	0.98	0	-	13/Peration+ Y score+ SN+
iter_size	CONST *	1	1	1		
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Folk me on cithub





• Every dataset, session and model have uniform resource identifier.

CIFAR_10

CIFAR_10/researcher_A/24

CIFAR_10/researcher_A/24/322

0	sagemaker-mxnet-py2-gpu-2017-11-26-22-30-36-301	Nov 26, 2017 22:30 UTC	_	() InProgre
0	sagemaker-mxnet-py2-cpu-2017-11-23-11-26-10-547	Nov 23, 2017 11:26 UTC	6 minutes	⊘ Complet
0	sagemaker-mxnet-py2-cpu-2017-11-23-11-15-08-321	Nov 23, 2017 11:15 UTC	6 minutes	⊗ Failed



URI

{Dataset} / {User id} / {Session id} / {Model id}

CIFAR 10 dataset

research_A's 24th session for CIFAR_10

Snapshot from epoch 322

855

Cloud ML Job, example_5_train_20170404_2... -

ted

Easy One-Liner CLI



Dataset registration

/app/examples/09_NMT\$ nsml dataset push NMT_EN_KR ./nmt_en_kr



Easy One-Liner CLI

Dataset registration

/app/examples/09_NMT\$ nsml dataset push NMT_EN_KR ./nmt_en_kr



/app/examples/09_NMT\$ nsml run -d NMT_EN_KR Session clair/NMT_EN_KR/1 is running



Easy One-Liner CLI

Dataset registration

/app/examples/09_NMT\$ nsml dataset push NMT_EN_KR ./nmt_en_kr



/app/examples/09_NMT\$ nsml run -d NMT_EN_KR Session clair/NMT_EN_KR/1 is running

Serve

/app/examples\$ echo Hello | nsml infer clair/NMT_EN_KR/1/12 안 녕 하 세 요



Easy One-Liner CLI

Parallel Experiments to Kill Slack

Distributed responses





[/app/examples/02_mnist\$ nsml run main.py -- --lr 0.1 Session KR18284/None/12 is running [/app/examples/02_mnist\$ nsml run main.py -- --lr 0.01 Session KR18284/None/13 is running [/app/examples/02_mnist\$ nsml run main.py -- --lr 0.001 Session KR18284/None/14 is running [/app/examples/02_mnist\$ nsml ps Name Created Args just now main.py --lr 0.001 KR18284/None/14 main.py --lr 0.01 just now KR18284/None/13 main.py --lr 0.1 KR18284/None/12 seconds ago



• Balance your brain to understand without effort



https://www.interaction-design.org/literature/book/the-encyclopedia-of-human-computer-interaction-2nd-ed/data-visualization-for-human-perception





Flexible Analysis

NSML





Visualization tool

nsml.report(

epoch=epoch+config.iteration, epoch_total=config.epochs, iter=iter_idx, iter_total=total_length, batch_size=batch_size, train_loss=running_loss / num_runs, train_accuracy=running_acc / num_runs, scope=locals()



examples - - bash - 90×53

07_vae_gan 08_LiteNet 09_movie_review



INE



examples - - bash - 90×53

07_vae_gan 08_LiteNet 09_movie_review



INE



TOGGLE ALL RUNS

/tmp/tensorboard





TOGGLE ALL RUNS

/tmp/tensorboard



Dynamic Control Flow

Typical training loop

Forward pass **Backward pass Communicate to NSML**

nsml.report(epoch=epoch+config.iteration, epoch_total=config.epochs, iter_idx, iter_total=total_length, batch_size=batch_size, train__loss=running_loss / num_runs, train__accuracy=running_acc / num_runs, scope=locals()





Terminal Graph

KR18284/mnist_torch/40

Train Epoch:	4	[1920/60000	(3%)] Loss:	0.294512 Au
Train Epoch:	4	[2560/60000	(4%)] Loss:	0.440668 Au
Train Epoch:	4	[3200/60000	(5%)] Loss:	0.419388 Au
Train Epoch:	4	[3840/60000	(6%)] Loss:	0.378146 Au
Train Epoch:	4	[4480/60000	(7%)] Loss:	0.398869 Au
Train Epoch:	4	[5120/60000	(9%)] Loss:	0.718725 Au
Train Epoch:	4	[5760/60000	(10%)] Loss	: 0.430032 A
Train Epoch:	4	[6400/60000	(11%)] Loss	: 0.363388 A
Train Epoch:	4	[7040/60000	(12%)] Loss	: 0.435897 A
Train Epoch:	4	[7680/60000	(13%)] Loss	: 0.313308 A
Train Epoch:	4	[8320/60000	(14%)] Loss	: 0.303983 A
Train Epoch:	4	[8960/60000	(15%)] Loss	: 0.378267 A

4

>>>



Terminal 🗘

1

- X LOSS: 0.0/1541
- x Loss: 0.126457
- x Loss: 0.078211
- x Loss: 0.067423
- x Loss: 0.082014
- x Loss: 0.170943
- ux Loss: 0.096298
- ux Loss: 0.083657
- ux Loss: 0.102749
- ux Loss: 0.091727
- ux Loss: 0.064584
- ux Loss: 0.097526

Terminal Graph

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- ux Loss: 0.097526

• Base of advanced features like save, load, infer, ...

```
[/tmp$ nsml run -d mnist_torch
Session KR18284/mnist_torch/12 is running
[/tmp$ nsml exec KR18284/mnist_torch/12 model
Net
  (conv1): Conv2d(1, 10, kernel_size=(5, 5), stride=(1, 1))
  (conv2): Conv2d(10, 20, kernel_size=(5, 5), stride=(1, 1))
  (conv2_drop): Dropout2d (p=0.5)
  (fc1): Linear (320 -> 50)
  (fc1_bn): BatchNorm1d(50, eps=1e-05, momentum=0.1, affine=True)
  (fc2): Linear (50 -> 10)
  (conv2_bn): BatchNorm2d(20, eps=1e-05, momentum=0.1, affine=True)
```



Bring Your Own Workspace

- (Almost) Nothing to learn
- Cached (Fast)



Bring Your Own Workspace

- (Almost) Nothing to learn
- Cached (Fast)



```
#nsml: nakosung/pytorch:latest-gpu-py3
from distutils.core import setup
setup(
    name='nsml example 07 VAE GAN',
    version='1.0',
    description='ns-ml',
    install_requires =[
        'visdom',
        'pillow'
```





No Framework Lock-in



tensorflow/tensorflow

By tensorflow • Free Under the Docker Co

Official docker images for deep learning f

Community



floydhub/pytorch

By floydhub · Free Under the Docker Com

pytorch

Community



gw000/keras

By gw000 · Free Under the Docker Comm

Keras in Docker for reproducible deep lea

Community



kaixhin/cuda-torch

By kaixhin • Free Under the Docker Comm

Ubuntu Core 14.04 + CUDA + Torch7 (inclu

Community



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Community License	
; framework TensorFlow (http://www.tensorflow.org)	
	± 8.3K
ommunity License	
	쇼 10K+
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earning on CPU or GPU	
	+ 5.04
nmunity License	5.0K ت
cluding iTorch).	

Interactive Mode

/app/examples/02_mnist\$ n



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Interactive Mode

/app/examples/02_mnist\$ n



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Pragmatic Research









Datasets > Sessions > Models



epoch



Datasets > Sessions > Models



epoch



Datasets > Sessions > Models

rev : 166

exited(0) · 2 days ago gpu : 1 args : main_category1.py --litenet --layer_config 6 6 6 -- use_dtcwt

cohn_kanade

Infer

SAVED MODEL · 160

epoch	160
reltime	8832.319140672684
test/err	1.046337817638266
test/loss	0.05320112881639079
test/top1	98.95366218236174
test/top5	99.85052316890882
total_epoch	160
walltime	1507716232.7313364

SAVED MODEL · 155

epoch	155
reltime	8545.214187383652
test/err	4.633781763826607
test/loss	0.2739498511914537
test/top1	95.3662182361734
test/top5	99.70104633781764
total_epoch	160
walltime	1507715945.6308603



SAVED MODEL · 150

epoch	150
reltime	8259.60128736496





Datasets > Sessions > Models

rev : 166

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cohn_kanade

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SAVED MODEL · 150

epoch	150
reltime	8259.60128736496





Datasets > Sessions > Models

 movie_review 19 hours ago · 606.08 MB 	= movie_review
✓ rev : 299	
exited(0) · 20 hours ago	
gpu : 1	
desc : partial data only(10000) - preprocess	sing
bug fixed	Input
SAVED MODEL · 998	기네아는 그 비아 하는
epoch 998	기대안아고 와아 일듯
reltime 6224.894388914108	
test/loss 6.172693252563477	
train/loss 0.19896139577031136	
walltime 1507819854.5514421	
SAVED MODEL · 996	
epoch 996	
reltime 6212.374230384827	
test/loss 6.237739562988281	
train/loss 0.19785388931632042	
walltime 1507819842.041744	
SAVED MODEL · 994	Auto ON
epoch 994	AULO UN
reltime 6199.914450883865	

6.1442670822143555

0.1978157013654709

test/loss

train/loss





Datasets > Sessions > Models

 movie_review 19 hours ago · 606.08 MB 	= movie_review
✓ rev : 299	
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6.1442670822143555

0.1978157013654709

test/loss

train/loss





Collaboration and Competition Leaderboard, CI-ML





New Workflow for ML Research

Collaboration and Competition Leaderboard, CI-ML





Collaborative Research

• Easy to reproduce and extend other's research.

/tmp\$





Collaborative Research

• Easy to reproduce and extend other's research.

/tmp\$





Cohesive and Competitive

Dataset-centric environment

Models are ranked automatically

Standardized and Quantified

Easy to compete Towards AutoML







Cohesive and Competitive

Dataset-centric environment

Models are ranked automatically

Standardized and Quantified

Easy to compete **Towards AutoML**













Quantitive model analysis makes ML workflow as a gym of AutoML



AutoML

Sota server

REST API

https://service.nsml.navercorp.com/ASR

\$ curl -X POST https://service.nsml.navercorp.com/ASR \
 -H "Content-Type: audio/wav" \
 --data-binary "@sample.wav"





Sota server Dataset ASR

REST API

https://service.nsml.navercorp.com/ASR

\$ curl -X POST https://service.nsml.navercorp.com/ASR \
 -H "Content-Type: audio/wav" \
 --data-binary "@sample.wav"













hank you

Several Hundreds of GPUs for this alpha (free)

https://research.clova.ai/nsml-alpha









