

Deep Learning Frameworks and Optimization Paths on Intel® Architecture

Andres Rodriguez – Solutions Architect, Intel Data Center Group Ravi Panchumarthy – Systems Engineer, Intel Data Center Group Elvis Jones – Solutions Architect, Amazon Web Services



Agenda

- Deep learning and frameworks overview
- Optimizations in Intel[®] Math Kernel Library (Intel[®] MKL) and Intel[®] MKL-DNN
- Intel optimized frameworks on Amazon Web Services (AWS*)



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Classification

Label the image

- Person
- Motorcyclist
- Bike

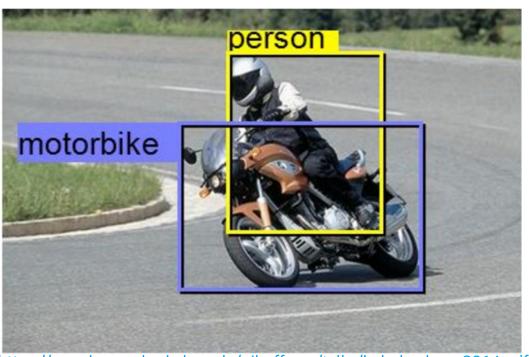


https://people.eecs.berkeley.edu/~jhoffman/talks/lsda-baylearn2014.pdf



Detection

Detect and label objects



https://people.eecs.berkeley.edu/~jhoffman/talks/lsda-baylearn2014.pdf



Semantic Segmentation

Label every pixel



https://people.eecs.berkeley.edu/~jhoffman/talks/lsda-baylearn2014.pdf



Natural Language Object Retrieval

a scene with three people query='man far right'





query='left guy'



query='cyclist'



http://arxiv.org/pdf/1511.04164v3.pdf



Visual and Textual Question Answering



What is the main color on the bus?



Answer: blue



What type of trees are in the background?



Answer: pine



How many pink flags are there?



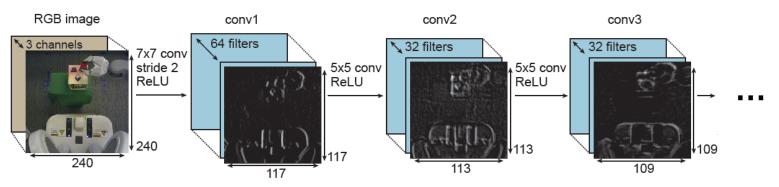
Answer: 2

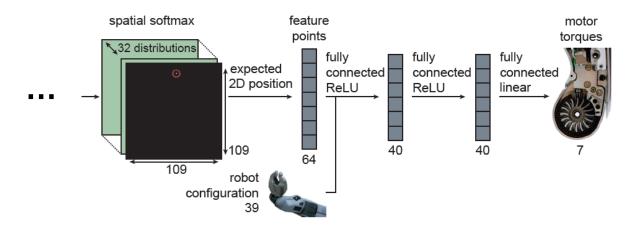


Is this in the wild?



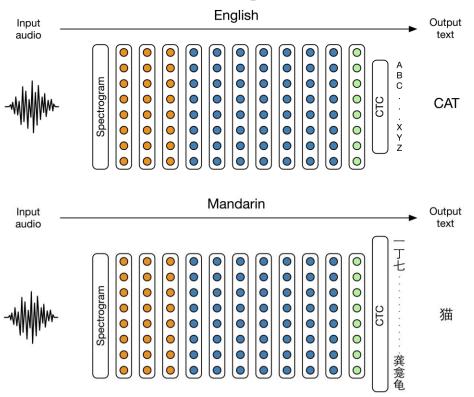
Visuomotor Control







Speech Recognition



The same architecture is used for English and Mandarin Chinese speech recognition

- Convolution Layer
- Recurrent Layer
- Fully Connected Layer



Q&A Natural Language Understanding

Question: Where was Mary before the Bedroom?

Answer: Cinema.

Facts	Episode 1	Episode 2	Episode 3
Yesterday Julie traveled to the school.			_
Yesterday Marie went to the cinema.			
This morning Julie traveled to the kitchen.			
Bill went back to the cinema yesterday.			
Mary went to the bedroom this morning.			
Julie went back to the bedroom this afternoon.			
[done reading]			





Personal Assistant









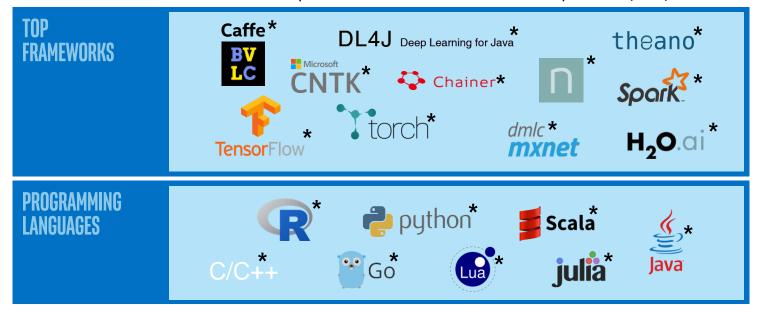




DL Tools

Machine Learning

Autonomous computation methods that learn from experience (data)



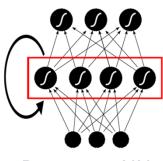


Agenda

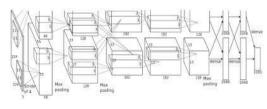
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Diversity in Deep Networks



Recurrent NN



CNN - AlexNet*



Variety in Network Topology

 Recurrent NNs common for NLP/ASR, DAG for GoogLeNet, Networks with memory...

But there are a few well defined building blocks

- Convolutions common for image recognition tasks
- GEMMs for recurrent network layers—could be sparse
- ReLU, tanh, softmax

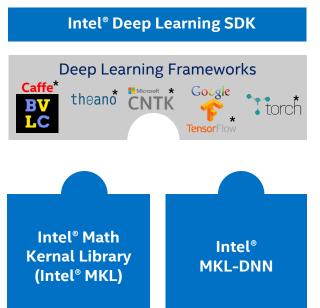


Intel® Math Kernel Library (Intel® MKL)

- Optimized AVX-2 and AVX-512 instructions
- Intel® Xeon® and Intel® Xeon Phi™ processors
- Supports all common layers types
- Coming soon: Winograd-based convolutions



Intel Deep Learning Software Stack



Intel® Xeon Phi™



Intel Deep Learning SDK – free tools to accelerate design, training and deployment of deep networks

Targeted release: Q4' 2016

Intel® MKL-DNN – free open source DNN functions designed for max Intel HW performance and high-velocity integration with DL frameworks



FPGA

Targeted release: Q4' 2016 (APIs and preview Q3' 2016)

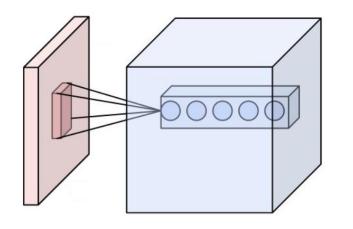
- Open source DNN functions included in MKL 2017
- IA optimizations contributed by community
- Binary GEMM functions
- Apache* 2 license

Intel libraries as path to bring optimized ML/DL frameworks to Intel hardware



Intel® Xeon®

Naïve Convolution



```
1: for i_0 \in 0, \ldots, minibatch do
2: for i_1 \in 0, \ldots, ifm do
3: for i_2 \in 0, \ldots, ofm do
4: for i_3 \in 0, \ldots, out_h do
5: for i_4 \in 0, \ldots, out_w do
6: for i_5 \in 0, \ldots, k_h do
7: for i_6 \in 0, \ldots, k_w do
8: output[i_0, i_1, i_3, i_4] + = input[i_0, i_1, i_3 * s + i_5 - 1, i_4 * s + i_6 - 1] * wts[i_1, i_2, i_5, i_6]
```

https://en.wikipedia.org/wiki/Convolutional_neural_network



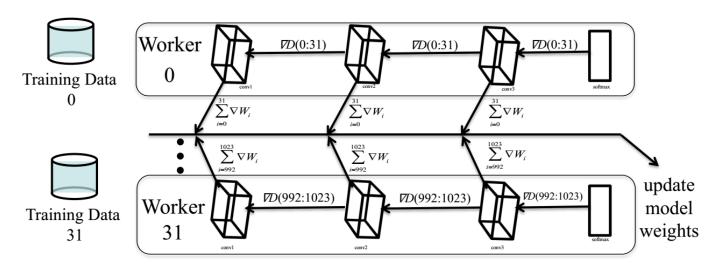
Cache Friendly Convolution

```
1: for i_0 \in 0, \ldots, minibatch do
                                                                                                      for i_8 \in 0, \ldots, RB_h do
                                                                          16:
       for i_1 \in 0, \ldots, ifm/SW do
                                                                                                         for i_9 \in 0, \ldots, RB_w do
 3:
          for i_2 \in 0, \ldots, ofm/SW do
                                                                          18:
                                                                                                            req = i_8 * RB_w + i_9
             for i_3 \in 0, \ldots, out_h/RB_h do
                                                                          19:
                                                                                                             out_u = i_3 * RB_h + i_8
                for i_4 \in 0, \ldots, out_w/RB_w do
                                                                          20:
                                                                                                            out_x = i_A * RB_w + i_Q
                   for rb_h \in 0, \ldots, RB_h do
                                                                          21:
                                                                                                            inp_y = out_y * stride + i_6 - 1
                      for rb_w \in 0, \ldots, RB_w do
                                                                                                            inp_x = out_x * stride + i_7 - 1
                                                                          22:
                         req = rb_h * RB_w + rb_w
                                                                          23:
                                                                                                             vout[reg]
                         out_y = i_3 * RB_h + rb_h
                                                                                                             VFMA(vout[reg],
10:
                         out_x = i_4 * RB_w + rb_w
                                                                                                             bcast(input[i_0][i_1][out_u][out_x][0]))
11:
                         vout[req]
                                                                                                             ,vwt)
                         LOAD(output[i_0][i_2][out_y][out_x])
                                                                                              for rb_h \in 0, \ldots, RB_h do
                                                                          24:
                   for i_5 \in 0, \ldots, SW do
                                                                                                 for rb_w \in 0, \ldots, RB_w do
13:
                      for i_6 \in kh_{start}, \dots, kh_{end} do
                                                                          26:
                                                                                                   req = rb_h * RB_w + rb_w
                         for i_7 \in kw_{start}, \dots, kw_{end} do
14:
                                                                                                    out_y = i_3 * RB_h + rb_h
                            vwt = LOAD(wts[i_1 * SW +
15:
                                                                          28:
                                                                                                    out_x = i_4 * RB_w + rb_w
                            i_5[i_2][i_6][i_7][0])
                                                                          29:
                                                                                                    STORE(vout[reg], output[i_0][i_2][out_y][out_x])
                            for i_8 \in 0, \ldots, RB_h do
16:
```



Caffe* Optimized for Intel® Architecture

- All the goodness of BVLC Caffe* +
 - Integrated with Intel® Math Kernel Library (Intel® MKL) 2017
 - Multi-node distributed training



Forrest landola, et al., "Scaling DNN Training on Intel Platforms." 2016



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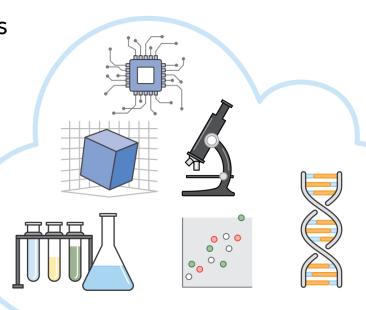


Unlimited infrastructure Low cost with flexible pricing Efficient clusters Why AWS* for HPC? Faster time to results Increased collaboration Concurrent Clusters on-demand



How is Amazon Web Services (AWS*) Used for HPC?

- High Performance Computing (HPC) for Engineering and Simulation
- High Throughput Computing (HTC) for Data-Intensive Analytics
- Collaborative Research Environments
- Monte-Carlo Simulations
- Data Visualization
- Hybrid Supercomputing centers
- Citizen Science
- Engineering/Science-as-a-Service
- Internet of Things (IOT)
- Serverless Computing



Goal: A simplified, repeatable, process

Goal = a simplified, repeatable, process for creating ML/DL clusters



CloudFormation* CfnCluster



Amazon Web Services (AWS*) CloudFormation*

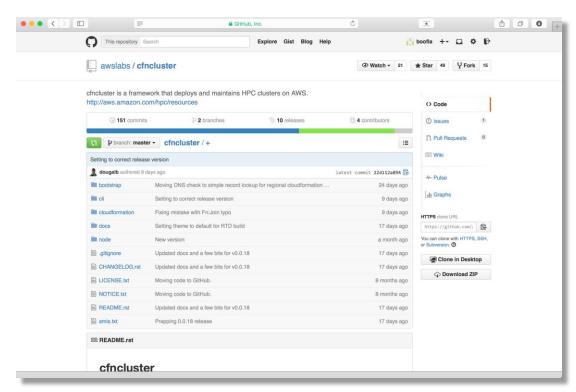


CloudFormation

- Fundamental service in AWS* used for automating deployment and configuration of resources
- CloudFormation* Template
 - JSON-formatted document which describes a configuration to be deployed in an AWS account
 - When deployed, refers to a "stack" of resources
 - Not a "script", a document



CfnCluster





https://aws.amazon.com/hpc/cfncluster https://github.com/awslabs/cfncluster



CfnCluster

- Made public 2014-06-10, as cfncluster-0.0.5
- Amazon* Software License https://aws.amazon.com/asl/
- Latest version is cfncluster-1.2.1, released on 2016-03-24



CfnCluster (cont)

- Architecture based on https://en.wikipedia.org/wiki/Beowulf_cluster
- OSes supported Amazon* Linux*, CentOS* 6 & 7, and Ubuntu* 1404
- Supports SGE, Openlava, Torque, and SLURM
- Extensible via Chef and pre/post-install scripts



Intel Optimized Frameworks on Amazon Web Services (AWS*)

Steps:

- Launch a CloudFormation* template
- 2. Edit the CfnCluster configuration file
- Use "cfncluster" to launch a cluster
- 4. Run some ML/DL examples



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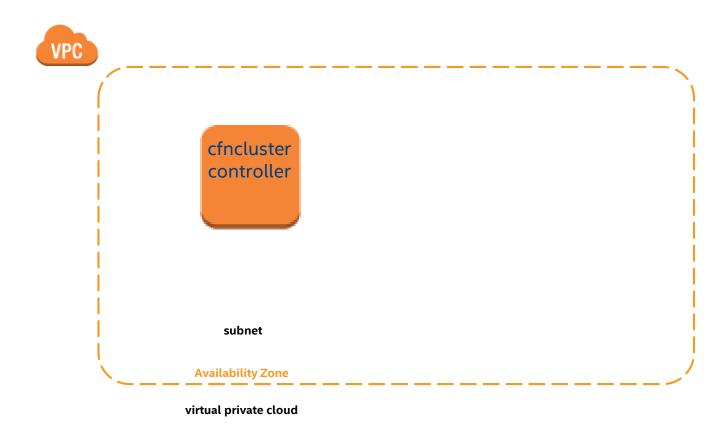


The CloudFormation stack





CloudFormation outputs:





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cfncluster configuration file details

```
[aws]
# The AWS region to run the cluster (i.e.: us-east-1, us-west-1, us-west-2, etc)
aws_region_name = us-west-2
# Set the following to the aws keys. If not defined, cnfcluster will attempt to use:
# a) environment variables
# b) an EC2 IAM role
aws_access_key_id = AWS_ACCESS_KEY
aws_secret_access_key = AWS_SECRET_ACCESS_KEY
[cluster default]
#create a name for your VPC
vpc settings = NAME OF THE VPC
key_name = NAME_OF_THE_SSH_KEY
# Override path to cloudformation in S3
# (defaults to https://s3.amazonaws.com/cfncluster-<aws_region_name>/templates/cfncluster-<version>.cfn.json)
# template_url = https://s3.amazonaws.com/cfncluster-us-east-1/templates/cfncluster.cfn.json
# the compute instance type
# (defaults to t2.micro)
compute instance type = c4.8xlarge
# the master instance type
# (defaults to t2.micro)
master_instance_type = c4.large
# Initial number of EC2 instances to launch as compute nodes in the cluster.
# (defaults to 2 for default template)
initial_queue_size = 2
# Maximum number of EC2 instances that can be launched in the cluster.
# (defaults to 10 for the default template)
max queue size = 2
# TD of a Custom AMT to use instead of published AMT's
```



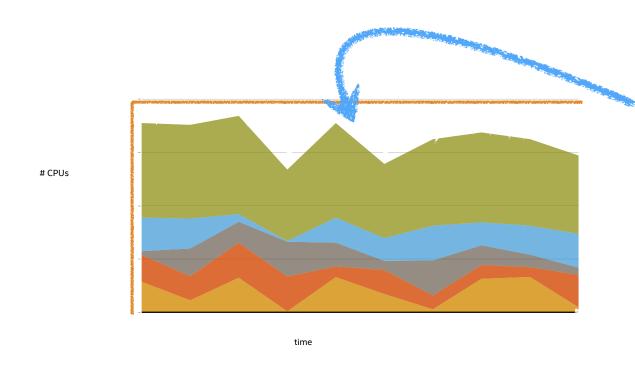
Config options to explore ...

Many options, but the most interesting ones immediately are:

```
# (defaults to t2.micro for default template)
compute instance type = c4.2xlarge
# Master Server EC2 instance type
# (defaults to t2.micro for default template
#master instance type = c4.4xlarge
# Inital number of EC2 instances to launch as co
                                                                       cluster.
# (defaults to 2 for default template)
                                                     Min & Max size of
#initial queue size = 0
                                                       vour cluster.
# Maximum number of EC2 instance
# (defaults to 10 for the default template)
#max queue size = 10
# Boolean flag to set autoscaling group to maintain
                                                                         scale back
                                                        Whether to fall
# (defaults to false for the default template)
                                                          back when
#maintain initial size = true
                                                        things get quiet
# Cluster scheduler
                                      Also can use
# (defaults to sge for the defau)
scheduler = sge
                                      'openlava' or
# Type of cluster to launch i.e.
                                        'torque'
# (defaults to ondemand for the default
                                                            Explore the SPOT
cluster type = spot
                                                           market if you want to
# Spot price for the ComputeFleet
                                                             save money :-)
spot price = 0.50
# Cluster placement group. This placement group must already exist.
# (defaults to NONE for the default template)
#placement group = NONE
                                                      A placement group will
                                                     provision your instances
                                                     very close to each other
                                                        on the network.
```



Amazon Web Services (AWS*) Spot Market



Spot Market

Our ultimate space filler.

Spot Instances allow you to name your own price for spare AWS* computing capacity.

Great for workloads that aren't time sensitive, and especially popular in research (hint: it's really cheap).



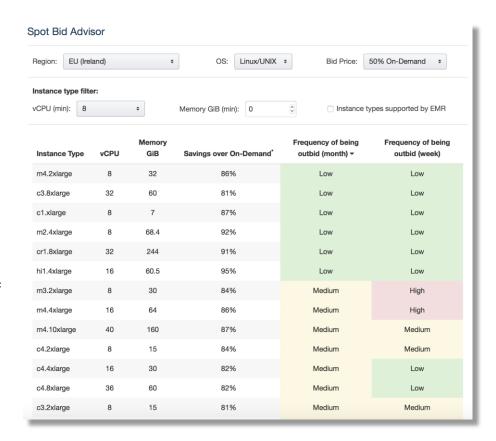
Spot Market - playing the [good] odds

Spot Bid Advisor

The Spot Bid Advisor analyzes Spot price history to help you determine a bid price that suits your needs.

You should weigh your application's tolerance for interruption and your cost saving goals when selecting a Spot instance and bid price.

The lower your frequency of being outbid, the longer your Spot instances are likely to run without interruption.





Intel Optimized Frameworks on AWS*

Steps:

- Launch a CloudFormation* template
- 2. Edit the CfnCluster configuration file
- 3. Use "cfncluster" to launch a cluster
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Components of our stack



Custom AMI

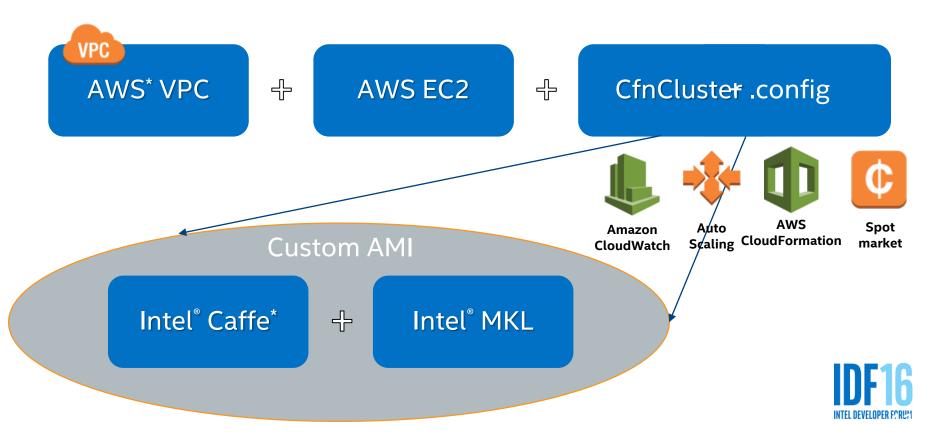


Components of our stack





Components of our stack



cfncluster command

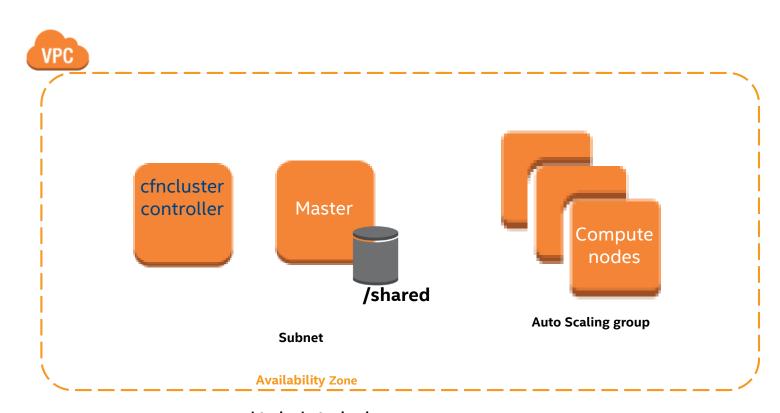
```
● ● AWS Credentials — ec2-user@ip-10-0-5-187:~ — ssh — 80×24

[ec2-user@ip-10-0-5-187 ~]$ cfncluster create myFirstCluster

Starting: myFirstCluster

Status: SNS — CREATE_IN_PROGRESS
```







Intel Optimized Frameworks on Amazon Web Services (AWS*)

Steps:

- Launch a CloudFormation* template
- 2. Edit the CfnCluster configuration file
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System-wide Upgrade from Intel® Xeon® v2 and Intel® Xeon® v3

```
$ ed ~/.cfncluster/config
/compute_instance_type/
compute_instance_type = c3.8xLarge
s/c3/c4/p
compute_instance_type = c4.8xLarge
w
949
$ cfncluster update boof-cluster
```

Yes, really :-)



This cluster intentionally left blank.

Your cluster is ephemeral.

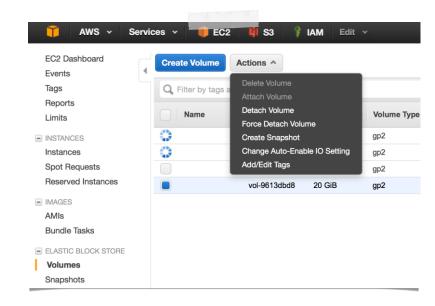
You've created a disposable cluster. But it's 100% recyclable It's worth noting that anything you put into this cluster will vaporize when you issue the command

\$ cfncluster delete <your cluster name>
... which might not be what you first expect.

It's easy to save your data and pick up from where you left off later

Before you delete your cluster, take a snapshot of the EBS (block storage) volume that you used for your /shared filesystem using the AWS* EC2 console (see the pic on the right)

The EBC volume you care most about is the one attached to the headnode instance (hint: it's probably the largest one)





Call to action

- Use Intel tools and optimized frameworks for DL workloads
 - https://software.intel.com/en-us/articles/training-and-deploying-deep-learning-networks-with-caffe-optimized-for-intel-architecture
 - https://software.intel.com/en-us/deep-learning-sdk
 - https://github.com/intelcaffe/caffe
 - https://github.com/intel/theano (other frameworks will be coming soon)
- Use multinode distributed training to reduce time-to-train
- Take advantage of AWS* CloudFormation* cluster
 - https://software.intel.com/en-us/articles/aws-cloudformation



Summary and Next Steps

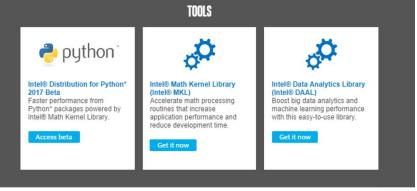
- Deep learning usages are expanding
- Intel and AWS* partnered to provide cloud users easy access to Intel optimized deep learning frameworks
- Reduce time to train by using Intel optimized frameworks and distribute the training workload across various nodes



Check out Intel® ML Developer Zone software.intel.com/machine-learning

- One developer portal for all Intel ML/DL tools, frameworks, training and support
- Download Caffe* Optimized for IA, Intel® MKL, Intel® DAAL, Intel® Distribution for Python*
- Community forums, articles, samples, tutorials and documentation







Processors Benefit

Machine Learning



Regression and

Classification



Applications using Intel®

Performance Libraries









A popular Python library designed to help write deep learning models.





Technical Sessions in Analytics Track

Tuesday, August 16, 2016

```
11:00 AM - 12:00 PM ANATS01 — Deep Learning Frameworks and Optimization Paths on Intel® Architecture Level 2 Room 2001
```

11:00 AM - 12:00 PM SOFTS01 — Accelerating Machine Learning on Apache Spark* Level 2 Room 2006

1:15 PM – 2:15 PM ANATSO3 — Enabling an End-to-End Architecture for Autonomous Cars Level 2 Room 2001

2:30 PM - 3:30 PM ANABIO1 — Advanced Analytics - Trends, Challenges, Opportunities Level 2 Room 2016 Tech & Business Insight

2:30 PM - 3:30 PM ANATS05 — How to Parallelize Neural Networks (xNNs) for Intel® Xeon Phi™ Level 2 Room 2001

4:00 PM - 5:00 PM ANATIO1 — Scaling to Meet the Growing Needs of Artificial Intelligence (AI) Level 2 Room 2016 Tech & Business Insight

4:00 PM – 5:00 PM ANATS07 — Innovative Use of Analytics and Machine Learning: Security, Network Function Virtualization (NFV), and Optimized Infrastructure Level 2 Room 2001

Wednesday, August 17, 2016

11:00 AM - 12:00 PM ANATS02 — Apache Spark* in Enterprise Analytics Level 2 Room 2002

1:15 PM - 2:15 PM ANATS04 — End-to-End Analytics Solutions with Trusted Analytics Platform Level 2 Room 2001

2:30 PM – 3:30 PM ANATS06 — The Complete Toolset for Accelerating Analytics – From Optimized System Architecture to Accelerators Level 2 Room 2001

2:30 PM – 3:30 PM IOTTS04 — IoT for Intervention During Equipment Failure Level 2 Room 2008

4:00 PM – 5:00 PM ANATS08 — Open Source Solutions for Network Intelligence Level 2 Room 2001



Experience Data Center Innovation & Get Engaged

Tech Showcase (1st Floor): see interactive demos & meet experts!

- Intel Pavilion : Cloud/SDI, Analytics, 5G technology demos
- Data Center communities: Artificial Intelligence, Intel Builders, NVMe, Memory
- Take part in the Data Center Solutions Tour for opportunity to win a drone!
- Meet experts in Intel Builders community for a chance to win an Intel NUC Kit!
- Live Data Center Q&A with Experts: 4-7 pm Wed in Networking Plaza

Data Center Experience Zone and Pentathlon (Floor 2 concourse)

Immerse yourself in data center-driven experiences in music, sports and more

Compete in five data center-related challenges based on Intel technologies to win a T-shirt and qualify to win awesome fabulous prizes!

- Tuesday: Rack Stack Challenge & Snap Telemetry Challenge, 11 a.m. 5 p.m.
- Wednesday: Silicon Photonics Challenge & Net Boss Challenge, 11 a.m. 5 p.m.
- Thursday: Masters Challenge @ 10:30 a.m., featuring top 3 scorers from Day 1 & 2 challenges



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- Software and workloads used in performance tests may have been optimized for performance only on Intel microprocessors. Performance tests, such as SYSmark and MobileMark, are measured using specific computer systems, components, software, operations and functions. Any change to any of those factors may cause the results to vary. You should consult other information and performance tests to assist you in fully evaluating your contemplated purchases, including the performance of that product when combined with other products. For more complete information visit http://www.intel.com/performance.
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