

INTEL OPENVINO & AZURE IOT EDGE

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Workshop Agenda

- Intel OpenVINO Overview:
 - Model Optimizer
 - Inference Engine
 - Pretrained Models & Demos
 - INT8 Calibration
- OpenVINO Release 2020.1 New Features
- OpenVINO & Azure IoTEdge Workflow
- OpenVINO & Azure IoTEdge Demo



INTEL® DISTRIBUTION OF OPENVINO™ TOOLKIT

Take your computer vision solutions to a new level with deep learning inference intelligence.



A toolkit to accelerate **high performance computer vision** & **deep learning inference into vision/AI applications** used from edge to cloud. It enables deep learning on hardware accelerators and easy deployment across multiple types of Intel® platforms.

Who needs this product?

- Computer vision, deep learning software developers
- Data scientists
- OEMs, ISVs, System Integrators

Usages

Security surveillance, robotics, retail, healthcare, AI, office automation, transportation, non-vision use cases (speech, text) & more.



HIGH PERFORMANCE, PERFORM AI AT THE EDGE



STREAMLINED & OPTIMIZED DEEP LEARNING INFERENCE



HETEROGENEOUS, CROSS-PLATFORM FLEXIBILITY

Free Download ▶ software.intel.com/openvino-toolkit

Open Source version ▶ 01.org/openvinotoolkit

Intel Computer Vision/AI Portfolio





















TOOLS

Intel® Parallel Studio XE Intel® System Studio Intel® Media SDK

Intel® Distribution of OpenVINO™ toolkit Intel® SDK for OpenCL[™] Applications **Nauta**

















Intel® Data **Analytics** Acceleration Library

Intel[®] Distribution for Python*

Intel® Math Kernel Library

Intel® nGraph™ Compiler

Movidius Stack













Compute









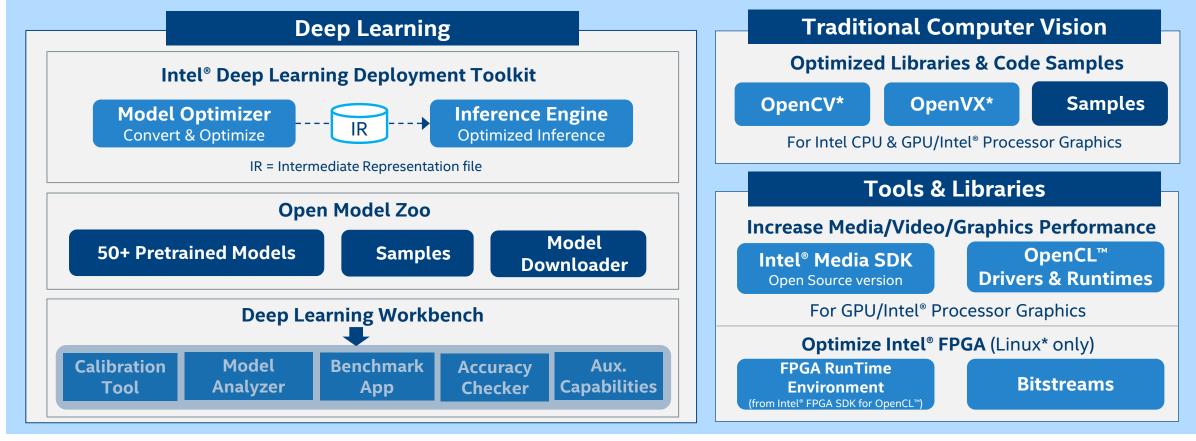






Visual Intelligence

What's Inside Intel® Distribution of OpenVINO™ toolkit



OS Support: CentOS* 7.4 (64 bit), Ubuntu* 16.04.3 LTS (64 bit), Microsoft Windows* 10 (64 bit), Yocto Project* version Poky Jethro v2.0.3 (64 bit), macOS* 10.13 & 10.14 (64 bit)

Intel® Architecture-Based Platforms Support















Intel® Vision Accelerator Design Products & Al in Production/ Developer Kits

An open source version is available at o1.org/openvinotoolkit (deep learning functions support for Intel CPU/GPU/NCS/GNA).



Intel® Deep Learning Deployment Toolkit

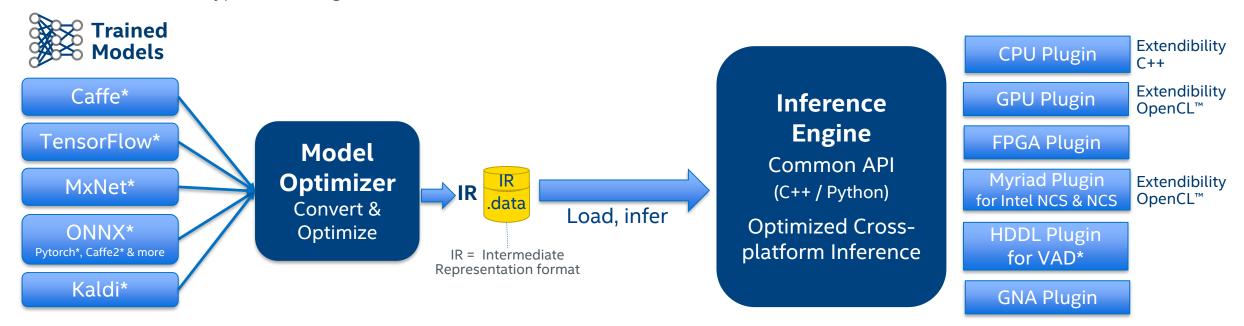
For Deep Learning Inference

Model Optimizer

- What it is: A Python*-based tool to import trained models and convert them to Intermediate representation.
- Why important: Optimizes for performance/space with conservative topology transformations; biggest boost is from conversion to data types matching hardware.

Inference Engine

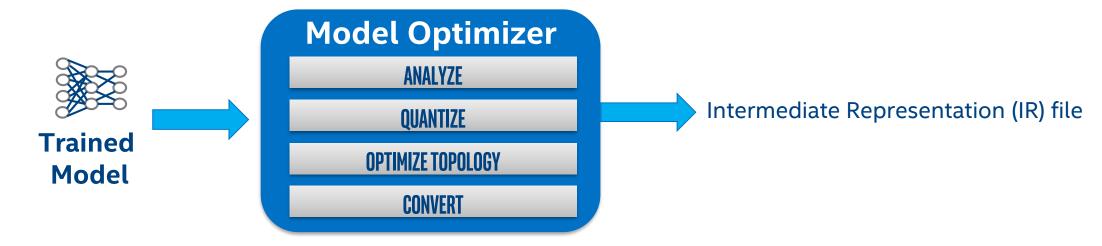
- What it is: High-level inference API
- Why important: Interface is implemented as dynamically loaded plugins for each hardware type. Delivers best performance for each type without requiring users to implement and maintain multiple code pathways.



GPU = Intel CPU with integrated GPU/Intel® Processor Graphics, Intel® NCS = Intel® Neural Compute Stick (VPU) *VAD = Intel® Vision Accelerator Design Products (HDDL-R)



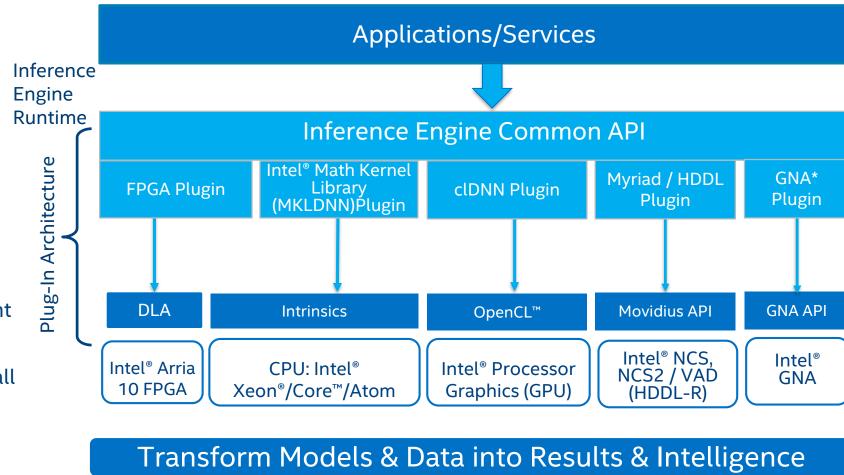
Improve Performance with Model Optimizer



- Easy to use, Python*-based workflow does not require rebuilding frameworks.
- Import Models from many supported frameworks: Caffe*, TensorFlow*, MXNet*, Kaldi*, exchange formats like ONNX* (Pytorch*, Caffe2* and others through ONNX).
- 100+ models for Caffe, MXNet, TensorFlow validated. Supports all ONNX* model zoo public models.
- Extends inferencing for non-vision networks with support of LSTM, Bert, GNMT, TDNN-LSTM, ESPNet and more.
- IR files for models using standard layers or user-provided custom layers do not require Caffe.
- Fallback to original framework is possible in cases of unsupported layers, but requires original framework.

Optimal Model Performance Using the Inference Engine

- Simple & unified API for inference across all Intel® architecture
- Optimized inference on large IA hardware targets (CPU/GEN/FPGA)
- Heterogeneity support allows execution of layers across hardware types
- Asynchronous execution improves performance
- Futureproof/scale your development for future Intel® processors
- Supports serialized FP16 IR across all plugins / platforms (CPU inference remains at FP32)



GPU = Intel CPU with integrated graphics/Intel® Processor Graphics/GEN GNA = Gaussian mixture model and Neural Network Accelerator



Speed Deployment with Pretrained Models & Samples

Expedite development, accelerate deep learning inference performance, speed production deployment

- Age & Gender
- Face Detection-standard & enhanced
- Head Position
- Human Detection—eye-level & high-angle detection
- Detect People, Vehicles & Bikes
- License Plate Detection: small & front facing
- Vehicle Metadata
- Human Pose Estimation
- Action recognition–encoder & decoder

- Text Detection & Recognition
- Vehicle Detection
- Retail Environment
- Pedestrian Detection
- Pedestrian & Vehicle Detection
- Person Attributes Recognition Crossroad
- Emotion Recognition
- Identify Someone from Different Videos–standard & enhanced
- Facial Landmarks
- Gaze estimation

- Identify Roadside objects
- Advanced Roadside Identification
- Person Detection & Action Recognition
- Person Re-identification—ultra small/ultra fast
- Face Re-identification
- Landmarks Regression
- Smart Classroom Use Cases
- Super Resolution
- Instance segmentation
- Image retrieval
- & more...

Binary Models

- Face Detection Binary
- Pedestrian Detection Binary

Vehicle Detection Binary

ResNet50 Binary

Speed Deployment with Demos

Expedite development, accelerate deep learning inference performance, speed production deployment

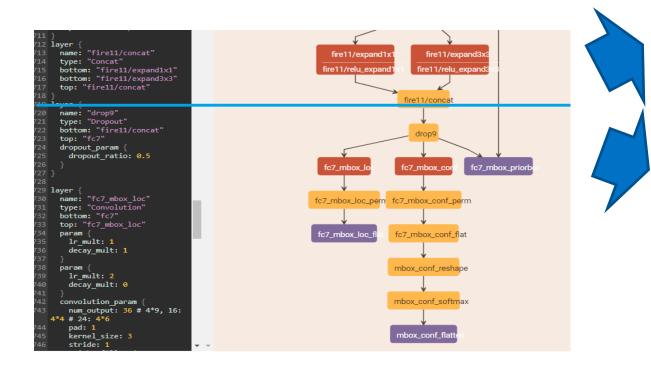
Some Available Demos in Intel® Distribution of OpenVINO™ toolkit

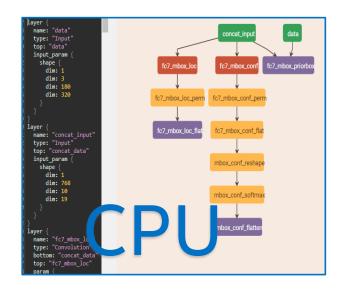
- Gaze/Pose Estimation
- Action Recognition
- Crossroad Camera
- Gaze Estimation
- Image Segmentation
- 3D Segmentation
- Instance Segmentation
- Interactive Face Recognition

- Text Detection & Recognition
- Mask R-CNN Object Detection
- Object Detection for Faster R-CNN
- Object Detection for SSD
- Object Detection for YOLO
- Super Resolution
- Text Detection
- Image Retrieval

Heterogeneous support

Possibility to execute different layers on different HW units



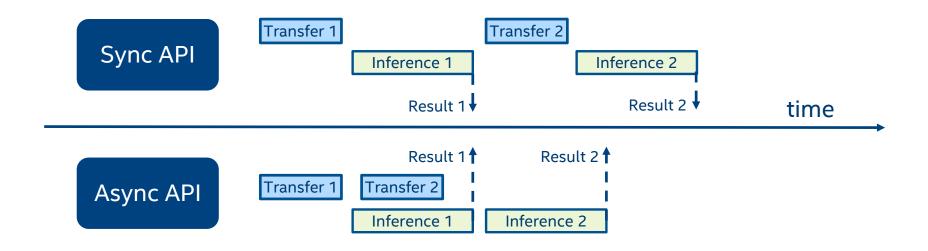




Synchronous vs Asynchronous execution

In IE API model executes by **Infer request** which can be:

- Synchronous blocks until inference is completed.
- Asynchronous checks the execution status with the wait, or specify a
 completion callback (recommended way).



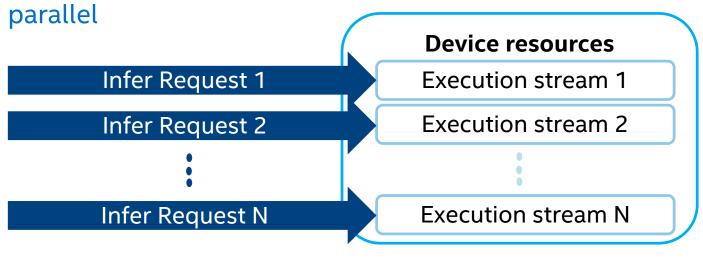
Inference Engine "Throughput" mode for CPU and iGPU

Latency – inference time of 1 frame (ms).

Throughput – overall amount of frames inferred per 1 second (FPS)

"Throughput" mode allows the Inference Engine to efficiently run multiple infer requests simultaneously, greatly improving the overall throughput.

Device resources are divided into execution "streams" – parts which runs infer requests in



time

Inference Engine Multi-Device Support

Automatic load-balancing between devices (inference requests level)

Fully general machinery: any combinations of devices

- CPU+iGPU
- Multiple NCS2, etc

As easy as "-d MULTI: HDDL, GPU" for cmd-line option of your favorite sample

C++ example (Python is similar)

```
// New IE-centric API
Core ie;
ExecutableNetwork exec = ie.LoadNetwork(network, {{"DEVICE_PRIORITIES", "HDDL,GPU"}}, "MULTI");
// Old plugin-centric API
auto plugin = PluginDispatcher().getPluginByDevice("MULTI:CPU,GPU");
ExecutableNetwork executable_network = plugin.LoadNetwork(network, config);
```

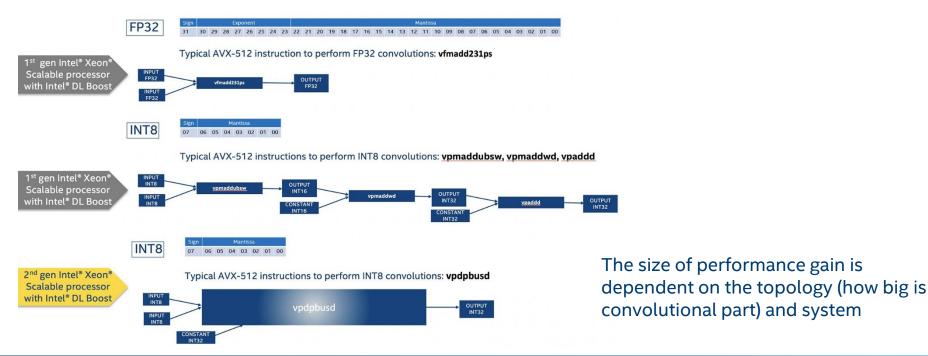
Int8 support on CPU – why important?

Significant performance boost and little loss of accuracy because

Benefit from less data size on Intel® platforms with Intel® AVX-512, Intel® AVX2, Intel® SSE4.2

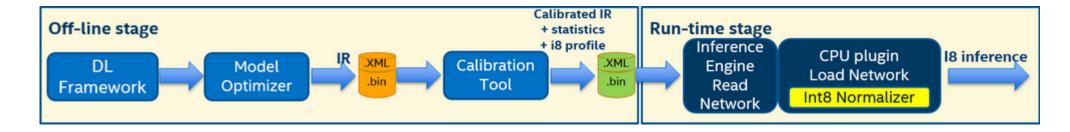
Take advantage from VNNI (Vector Neural Network Instructions) on 2nd Generation Intel®

Xeon® Scalable



Int8 Calibration – Calibration Tool

Calibration tool - command line app which collects statistics from FP32 or FP16 IR (intermediate representations)



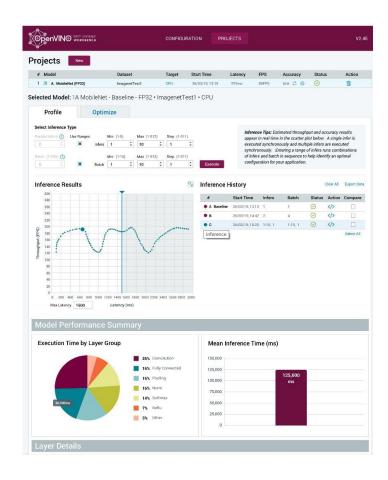
Calibration in "simplified" mode – see the maximum of potential performance gain from Int8 without accuracy calculation

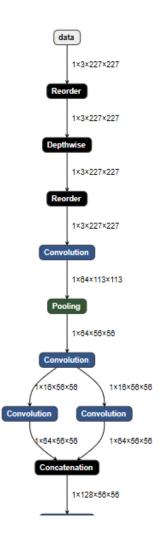
Pass full calibration process to get working Int8 model with accuracy statistics

Deep Learning Workbench

Deep Learning Workbench capabilities

- Web-based tool UI extension of Intel® Distribution of OpenVINO™ toolkit functionality
- Visualizes performance data for topologies/ layers to aid in model analysis
- Automate analysis for optimal performance configuration (streams, batches, latency)
- Experiment with int8 calibration for optimal tuning
- Provide accuracy info through accuracy checker
- Direct access to Models from public set of Open Model Zoo





OpenVINO™ Toolkit

OpenVINO

Open Source Version

- Provides flexibility and availability to the developer community to extend OpenVINO™ toolkit for custom needs
- Components that are open sourced
 - Deep Learning Deployment Toolkit with CPU, GPU, Heterogeneous, Myriad (for Intel® Neural Compute Stick (Intel® NCS) & Intel® NCS2), and GNA plugins
 github.com/opency/dldt
 - Open Model Zoo Includes pretrained models, model downloader, demos and samples: github.com/opency/open_model_zoo
- See <u>FAQ</u> and next slides for key differences between the open source and Intel distribution

Learn More ▶ <u>01.org/openvinotoolkit</u>



OpenVINO™ Toolkit

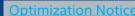
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 github.com/opency/dldt
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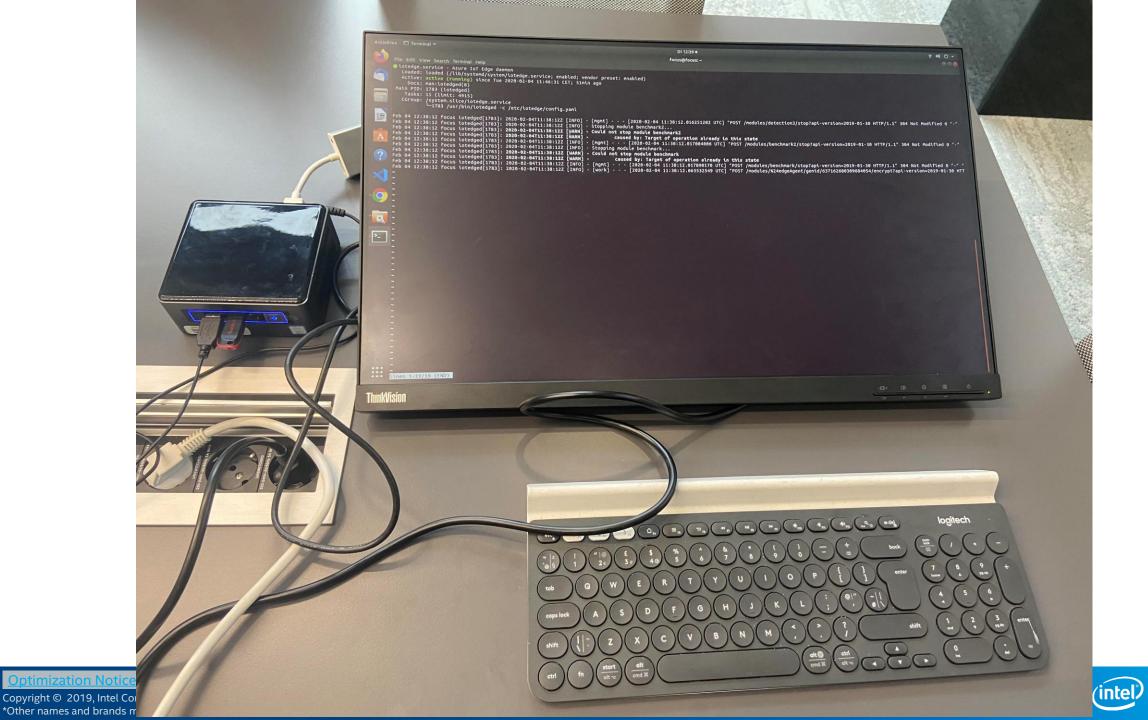
OpenVINO™ R2020.1 New Features



- Ngraph integration into OpenVINO
- Microsoft Visual Studio 2017/2019 compiler support
- Deployment Package Manager Windows
- New Int8 runtime –Post Training Optimization Tool (POT)
- New NUMA support (Multi-socket support for Windows
- Inference Engine C API
- Speech Libraries and End-to-End Speech Demos
- Inference & Streaming in OpenCV G-API
- 3D convolution with int8



OPENVINO & AZURE IOT EDGE DEMO



Workflow

1

Convert Model to IR



model .bin model .xml

2

Create OpenVINO Docker Image



Create Container Registry



Push Image to Container Registry

3

Create an IoT Hub



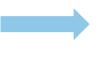
Register an IoT Edge device



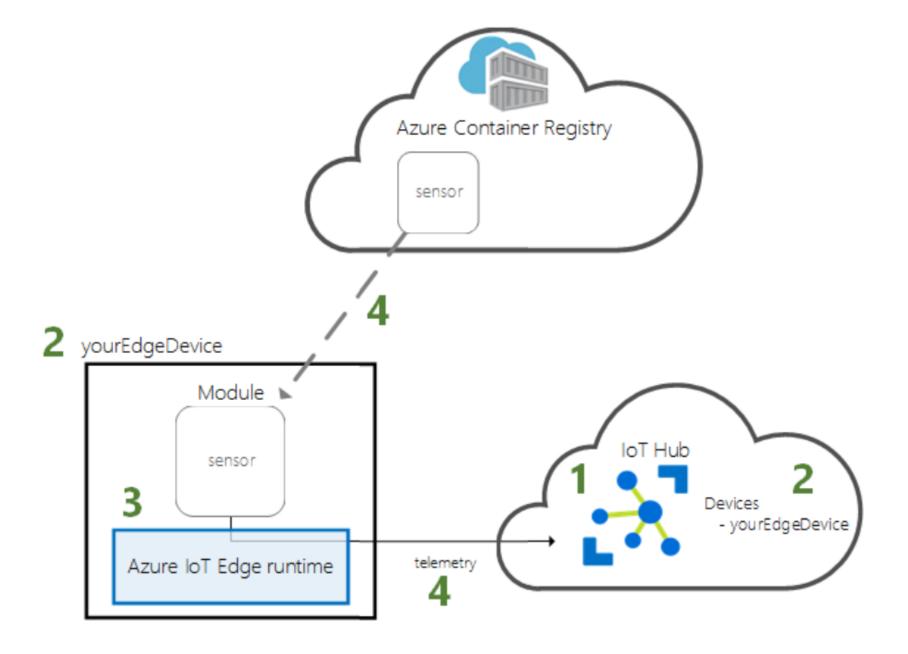
Configure Azure IoT Edge Runtime on Edge Device

4

Deploy Faas Module on Edge Device



Analytics





Create OpenVINO Docker Image

```
ubuntu:16.04
ENV http_proxy $HTTP_PROXY
ENV https proxy $HTTPS PROXY
ARG DOWNLOAD_LINK=http://registrationcenter-download.intel.com/akdlm/irc_nas/13231/l_openvino_toolkit_p_2019.0.000.tgz
ARG INSTALL DIR=/opt/intel/openvino
ARG TEMP_DIR=/tmp/openvino_installer
RUN apt-get update && apt-get install -y --no-install-recommends \
   wget \
   cpio \
   sudo \
   lsb-release && \
   rm -rf /var/lib/apt/lists/*
RUN mkdir -p $TEMP DIR && cd $TEMP DIR && \
   wget -c $DOWNLOAD_LINK && \
   tar xf l_openvino_toolkit*.tgz && √
   cd l_openvino_toolkit* && \
   sed -i 's/decline/accept/g' silent.cfg && \
    ./install.sh -s silent.cfg && \
   rm -rf $TEMP DIR
RUN $INSTALL DIR/install dependencies/install openvino dependencies.sh
 build Inference Engine samples
RUN mkdir $INSTALL_DIR/deployment_tools/inference_engine/samples/build && \
   cd $INSTALL DIR/deployment tools/inference engine/samples/build && \
   /bin/bash -c "source $INSTALL_DIR/bin/setupvars.sh && cmake .. && make -j1"
```

<u>Optimization Notice</u>



Create OpenVINO Docker Image

- https://docs.openvinotoolkit.org/latest/_docs_install_guides_installing_openvino_docker_linux.html
- Implement additional dependencies for building GPU, Movidius NCS and FPGA images

```
COPY intel-opencl*.deb /opt/gfx/
RUN cd /opt/gfx && \
    dpkg -i intel-opencl*.deb && \
    ldconfig && \
    rm -rf /opt/gfx
RUN useradd -G video -ms /bin/bash user

USER user
```

```
RUN cd /tmp/ && \
   wget https://github.com/libusb/libusb/archive/v1.0.22.zip && \
   unzip v1.0.22.zip && cd libusb-1.0.22 && \
   ./bootstrap.sh && \
   ./configure --disable-udev --enable-shared && \
   make -j4 && make install && \
   rm -rf /tmp/*
```

```
ENV CL_CONTEXT_COMPILER_MODE_INTELFPGA=3
ENV DLA_AOCX=/opt/intel/openvino/a10_devkit_bitstreams/2-0-1_RC_FP11_Generic.aocx
ENV PATH=/opt/altera/aocl-pro-rte/aclrte-linux64/bin:$PATH
```



Create and Push OpenVINO Docker Image

- docker build -t dockerimage.
- docker login openvinoregistry.azurecr.io -u XXXXX -p XXXXXX
- docker tag dockerimage openvinoregistry.azurecr.io.azurecr.io/dockerimage:v1
- docker push openvinoregistry.azurecr.io/dockerimage:v1



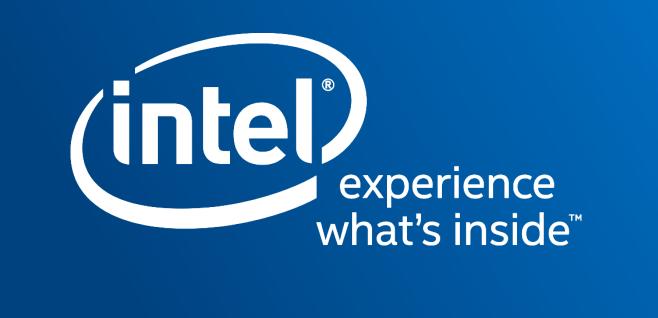
Configure Module on IoT Hub

- Create IoT Hub. Add Device. Set Connection String on Edge Device(/etc/iotedge/config.yaml)
- Set Device Module Image URI to Repo in Container Registry
- Set Device Module Container Create Options



OPENVINO & AZURE IOT EDGE DEMO

QNA



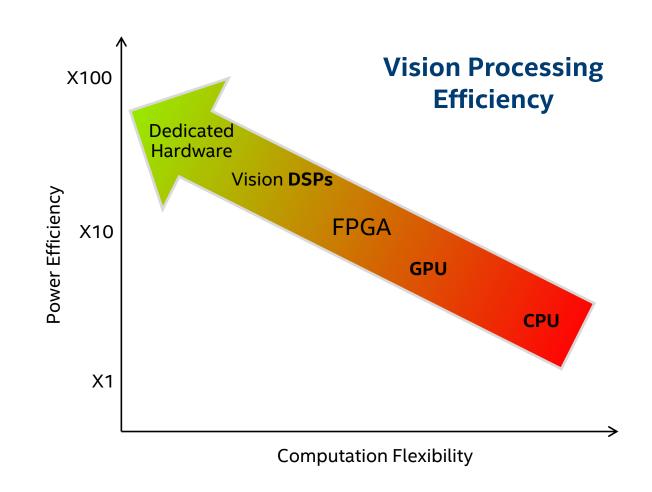
Choosing the "Right" Hardware

Power/Performance Efficiency Varies

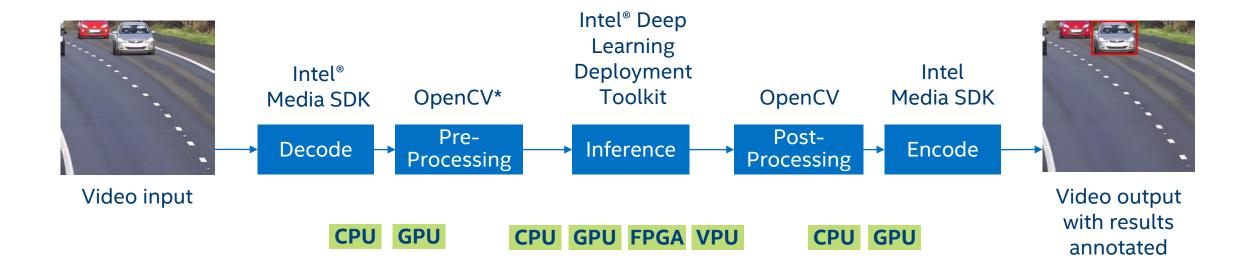
- Running the right workload on the right piece of hardware → higher efficiency
- Hardware acceleration is a must
- Heterogeneous computing?

Tradeoffs

- Power/performance
- Price
- Software flexibility, portability



End-to-End Vision Workflow



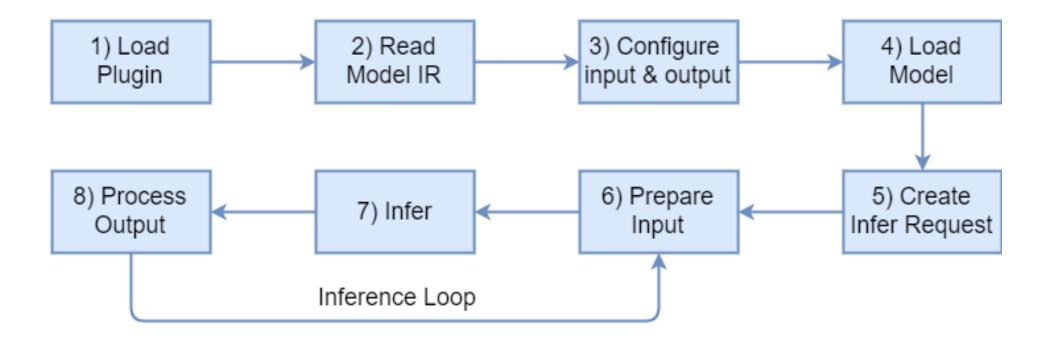
Model Optimizer

Model optimizer performs generic optimization:

- Node merging
- Horizontal fusion
- Batch normalization to scale shift
- Fold scale shift with convolution
- Drop unused layers (dropout)
- FP16/Int8 quantization
- Model optimizer can add normalization and mean operations, so some preprocessing is 'added' to the IR
 - --mean_values (104.006, 116.66, 122.67)
 - --scale values (0.07, 0.075, 0.084)

Hardware	FP32	FP16	FP11	INT8
CPU	yes	yes	no	yes
GPU	yes	yes	no	no
MYRIAD	no	yes	no	no
FPGA/DLA	no	yes	yes	no

Application Workflow for Inference Engine

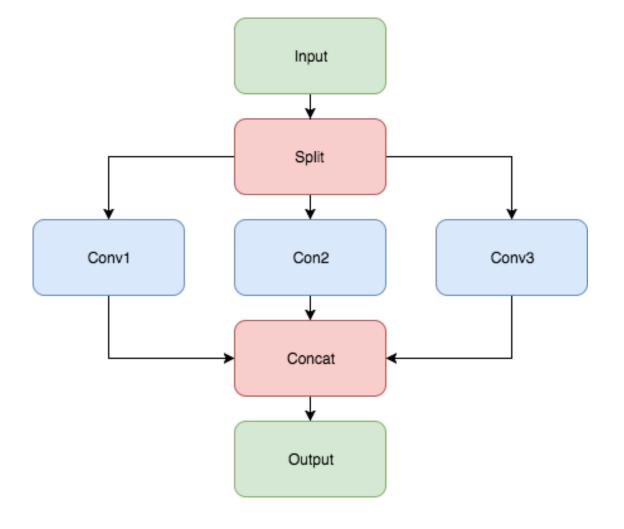




Internal CPU Plugin Optimizations

Merging of group convolutions.

- It means that if a topology contains the following pipeline →
- CPU plugin will merge it into one Convolution with the group parameter (Convolutions should have the same parameters).



Internal CPU Plugin Optimizations

 Fusing Convolution with ReLU or ELU. CPU plugin is fusing all Convolution with ReLU or ELU layers if these layers are located after the Convolution layer.

 Fusing Convolution + Sum or Convolution + Sum + ReLu. To improve performance, the CPU plugin fuses the following structure:

