

Choose the Best Accelerated Technology

# Intel® Distribution of OpenVINO™ toolkit

LRZ AI Workshop

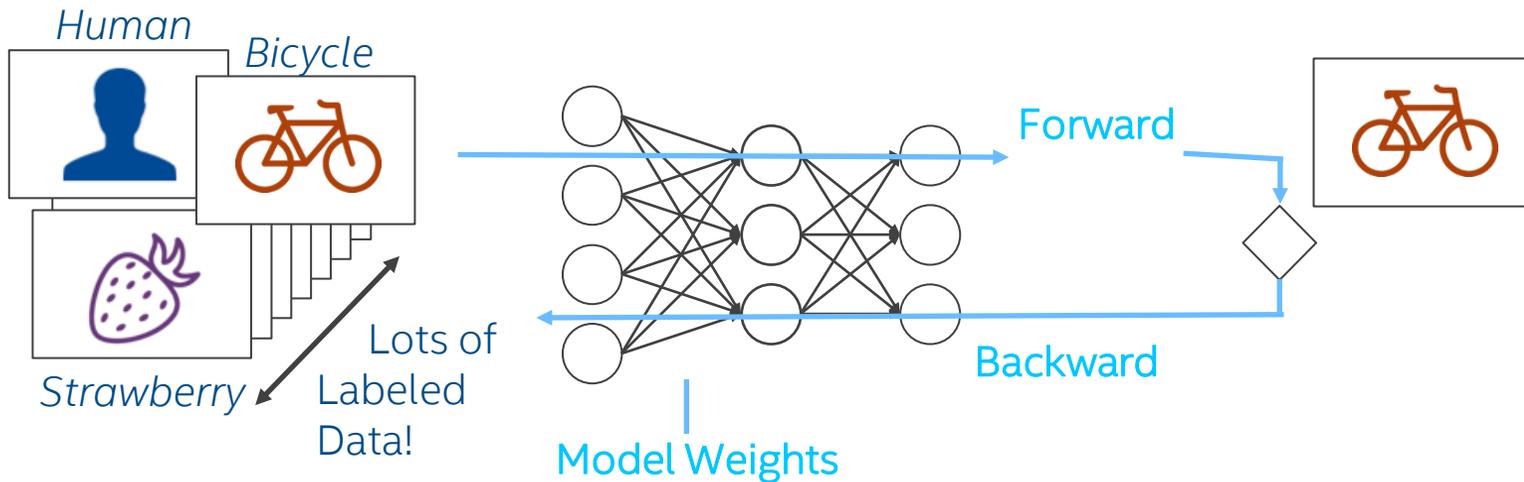
Roy Allela – AI Software Solutions Engineer

21.04.2022

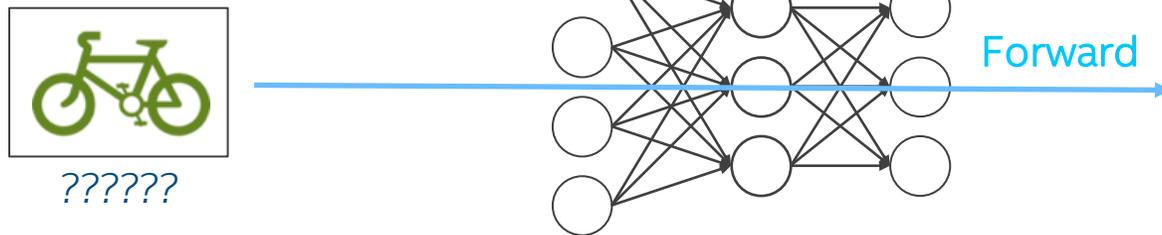


# Deep Learning: Training vs. Inference

## Training

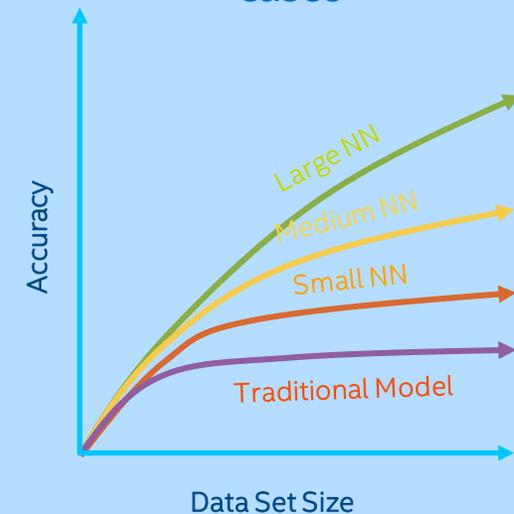


## Inference



## Did You Know?

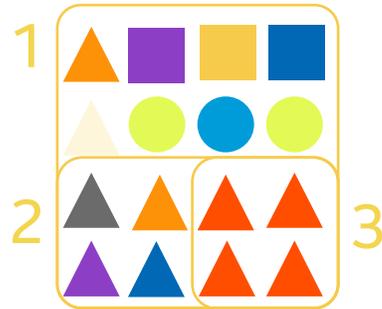
Training requires a very large data set and deep neural network (many layers) to achieve the highest accuracy in most cases



# AI Compute Considerations

How do you determine the right computing for your AI needs?

## WORKLOADS



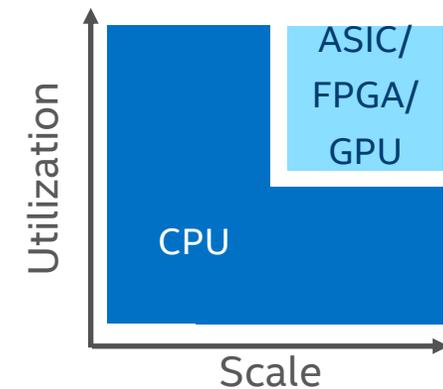
What is my workload profile?

## REQUIREMENTS



What are my use case requirements?

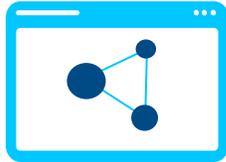
## DEMAND



How prevalent is AI in my environment?

# Challenges in Deep Learning

Development and deployment challenges in deep learning



## Unique Inference Needs

Gap in performance and accuracy between trained and deployed models

Low performing, lower accuracy models deployed



## Integration Challenges

No streamlined way for end-to-end development workflow

Slow time-to-solution and time-to-market

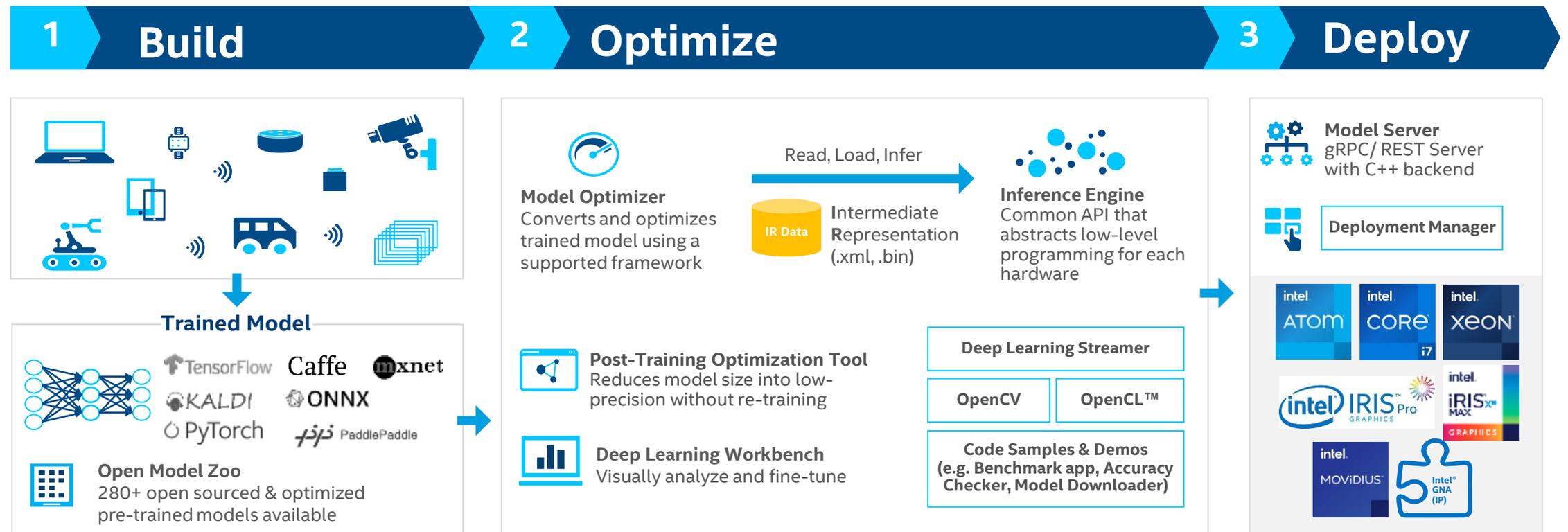


## No One Size Fits All

Diverse requirements for myriad use cases require unique approaches

Inability to meet use-case specific requirements

# Three steps for the Intel® Distribution of OpenVINO™ toolkit



# Supported Frameworks

Breadth of supported frameworks to enable developers with flexibility

011010110110  
110101101011  
001011010100  
011010110110  
110101101011  
001011010100  
011010110110  
110101101011  
001011010100  
011010110110  
110101101011  
001011010100  
011010110110  
110101101011  
001011010100  
011010110110  
110101101011  
001011010100  
011010110110  
110101101011  
001011010100



Trained Model



mxnet



Caffe



OpenVINO™



(and other tools via ONNX\* conversion)



LibSVM



SIEMENS



Supported Frameworks and Formats ▶ [https://docs.openvino toolkit.org/latest/\\_docs\\_IE\\_DG\\_Introduction.html#SupportedFW](https://docs.openvino toolkit.org/latest/_docs_IE_DG_Introduction.html#SupportedFW)  
Configure the Model Optimizer for your Framework ▶ [https://docs.openvino toolkit.org/latest/\\_docs\\_MO\\_DG\\_prepare\\_model\\_Config\\_Model\\_Optimizer.html](https://docs.openvino toolkit.org/latest/_docs_MO_DG_prepare_model_Config_Model_Optimizer.html)

# Model Optimization

Breadth of supported frameworks to enable developers with flexibility

**Model Optimizer** loads a model into memory, reads it, builds the internal representation of the model, optimizes it, and produces the **Intermediate Representation**.

Optimization techniques available are:

- Linear operation fusing
- Stride optimizations
- Group convolutions fusing

Note: Except for ONNX (.onnx model formats), all models have to be converted to an IR format to use as input to the Inference Engine



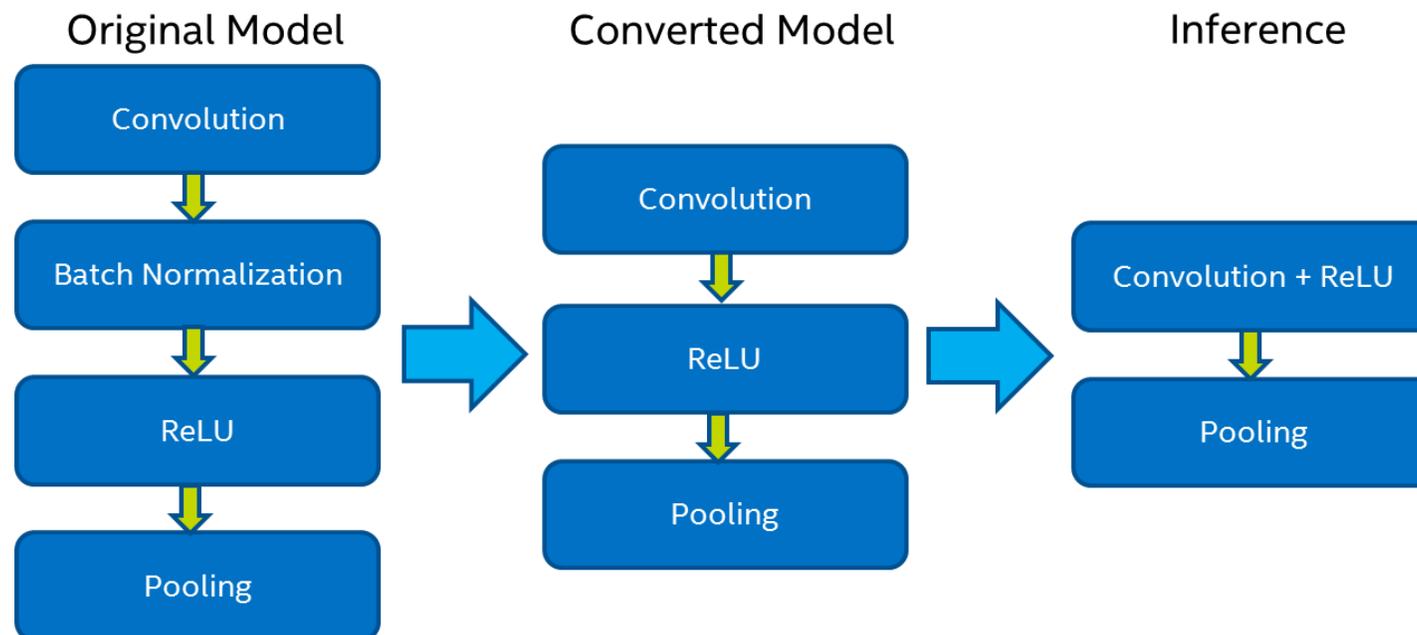
**.xml** – describes the network topology

**.bin** – describes the weights and biases binary data

# Model Optimizer: Linear Operation Fusing

## ■ Example

1. Remove Batch normalization stage.
2. Recalculate the weights to 'include' the operation.
3. Merge Convolution and ReLU into one optimized kernel.



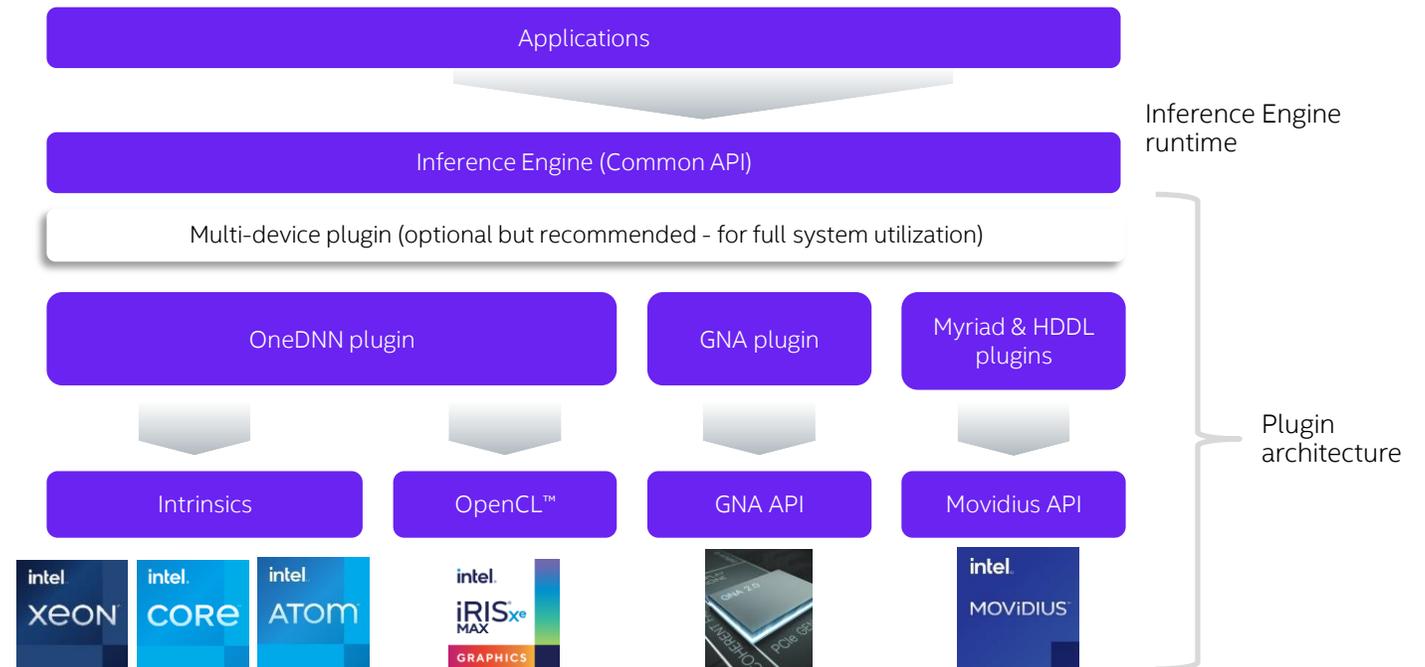
# Optimal Model Performance Using the Inference Engine

## Core Inference Engine Libraries

- Create Inference Engine Core object to work with devices
- Read the network
- Manipulate network information
- Execute and pass inputs and outputs

## Device-specific Plugin Libraries

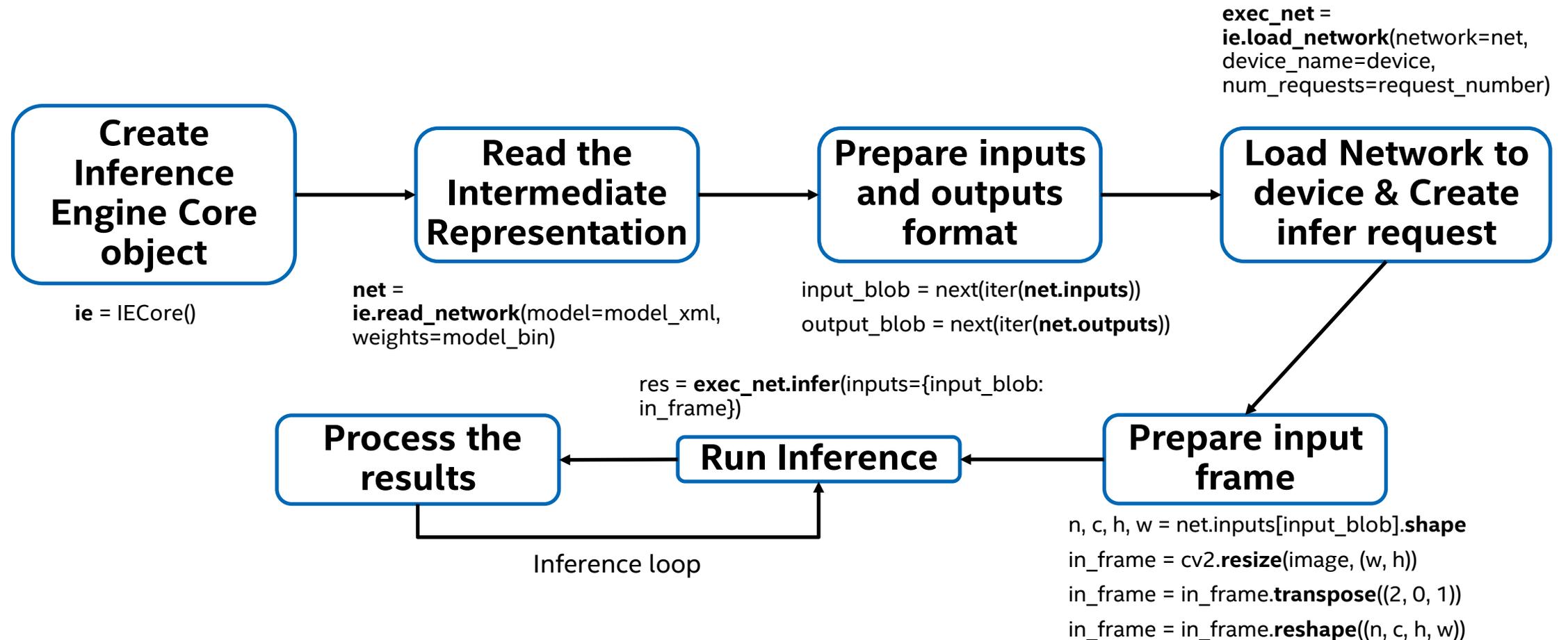
- For each supported target device, Inference Engine provides a plugin — a DLL/shared library that contains complete implementation for inference on this device.



GPU = Intel CPU with integrated graphics/Intel® Processor Graphics/GEN

GNA = Gaussian mixture model and Neural Network Accelerator

# Common Workflow for Using the Inference Engine API

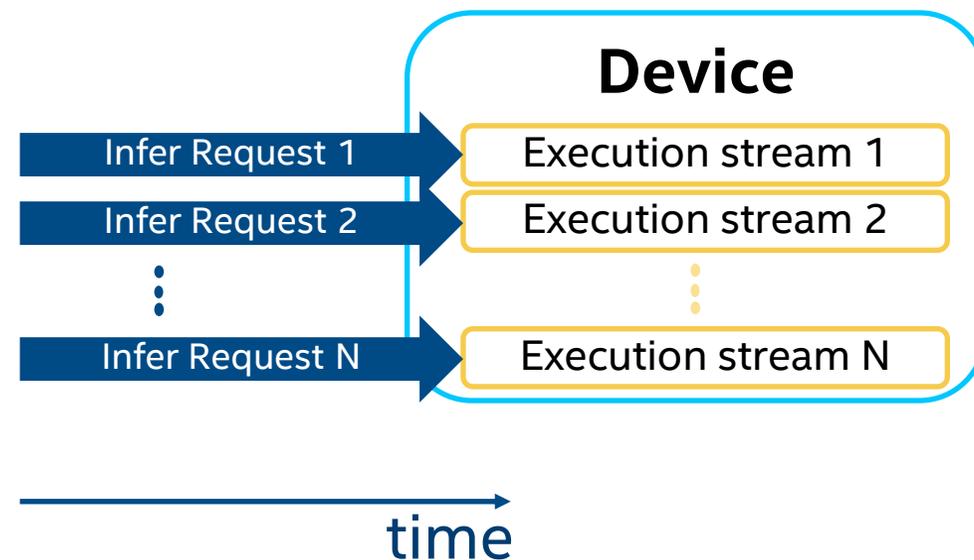


[http://docs.openvino toolkit.org/latest/docs\\_IE\\_DG\\_Integrate\\_with\\_customer\\_application\\_new\\_API.html](http://docs.openvino toolkit.org/latest/docs_IE_DG_Integrate_with_customer_application_new_API.html)

# Inference Engine

## Throughput Mode for CPU, iGPU and VPU

- 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 parallel



### CPU Example:

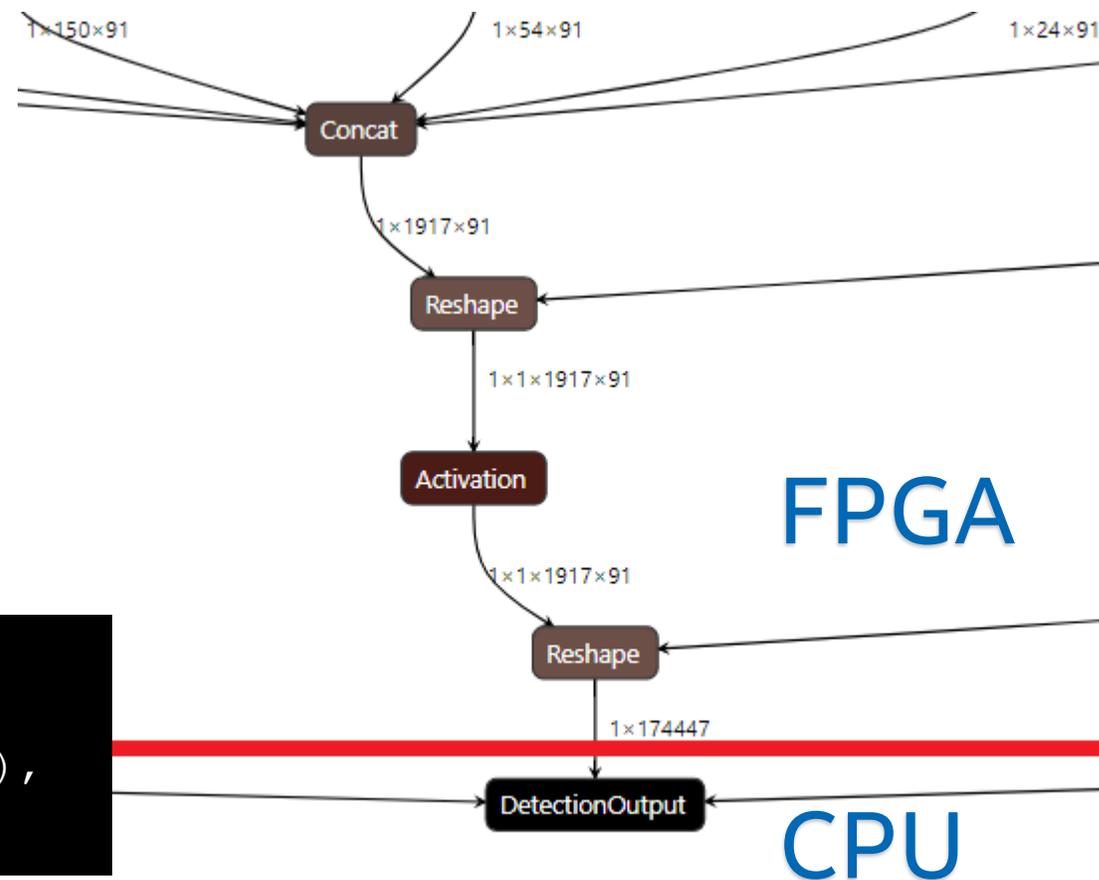
```
ie = IECore()  
ie.GetConfig(CPU, KEY_CPU_THROUGHPUT_STREAMS)
```

# Inference Engine

## Heterogeneous Support

- You can execute different layers on different HW units
- Offload unsupported layers on fallback devices:
  - Default affinity policy
  - Setting affinity manually (`CNNLayer::affinity`)
- All device combinations are supported (CPU, GPU, FPGA, MYRIAD, HDDL)
- Samples/demos usage “-d HETERO:FPGA,CPU”

```
InferenceEngine::Core core;  
auto executable_network =  
core.LoadNetwork(reader.getNetwork(),  
"HETERO:FPGA,CPU");
```

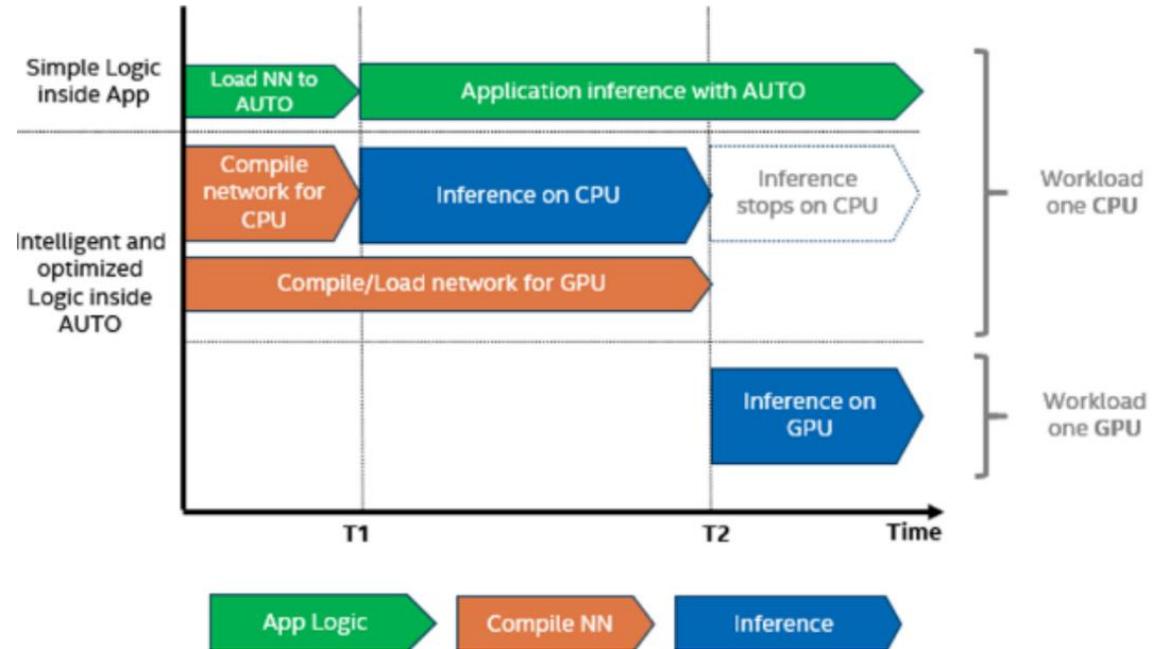


# Inference Engine

## Automatic Device Selection: AUTO

- AUTO discovers the accelerators & capabilities in the system, chooses and configures the device matching to the NN's precision, topology, user's performance requirements.
- Application portability, HW agnostic, dispatching workload to best fit device (CPU, GPU, VPU).

- AUTO always chooses the best device, if compiling model fails on this device, AUTO will try to compile it on next best device until one of them succeeds. If priority list is set, AUTO only select devices according to the list.



```
ov::CompiledModel model0 = core.compile_model(model);  
ov::CompiledModel model1 = core.compile_model(model, "AUTO");  
ov::CompiledModel model2 = core.compile_model(model, "AUTO", {});
```

# OpenVINO as execution provider

- You can use OpenVINO Inference Engine as backend of other DL Inference Frameworks such as Tensorflow or ONNX Runtime



- Benefit: the advantages of OpenVINO (multiple HW support and acceleration) in your favorite framework

# OpenVINO™ Integration with TensorFlow\*

Only 2 lines  
to be added  
to regular  
Tensorflow

```
1 # Installation steps
2 # more details : https://github.com/openvinotoolkit/openvino\_tensorflow
3 #pip3 install -U pip==21.0.1
4 #pip3 install -U tensorflow==2.4.1
5 #pip3 install openvino-tensorflow
6
7 # Import package and set backend
8 import openvino_tensorflow
9 openvino_tensorflow.set_backend('GPU')
10
11 # Load a TF Saved Model
12 model = tf.keras.models.load_model('resnet50_saved_model')
13
14 # Get the input size of the model
15 network_input_size = saved_model_loaded.input.shape()
16
17 # Resize the input image
18 resized_image = resize(input_image, network_input_size)
19
20 # Run inference
21 model.predict(resized_image)
```

CPU  
GPU  
MYRIAD  
VAD-M

# Pre-trained Models

Open-sourced repository of pre-trained models and support for public models



## Intel Pre-trained Models

[Action Recognition Models](#)  
[Classification Models](#)  
[Head Pose Estimation Models](#)  
[Human Pose Estimation Models](#)  
[Image Processing Models](#)

[Instance Segmentation Models](#)  
[Machine Translation Models](#)  
[Object Attribute Estimation Models](#)  
[Object Detection Models](#)  
[Optical Character Recognition](#)

[Models](#)  
[Question Answering Models](#)  
[Semantic Segmentation Models](#)  
[Text-to-speech Models](#)  
[Token Recognition Models](#)



## Public Pre-trained Models

[Action Recognition Models](#)  
[Classification Models](#)  
[Colorization Models](#)  
[Face Recognition Models](#)  
[Human Pose Estimation Models](#)  
[Image Inpainting Models](#)  
[Image Processing Models](#)  
[Image Translation Models](#)

[Instance Segmentation Models](#)  
[Monocular Depth Estimation Models](#)  
[Object Attribute Estimation Models](#)  
[Object Detection Models](#)  
[Optical Character Recognition Models](#)

[Place Recognition Models](#)  
[Semantic Segmentation Models](#)  
[Sound Classification Models](#)  
[Speech Recognition Models](#)  
[Style Transfer Models](#)  
[Text-to-speech Models](#)

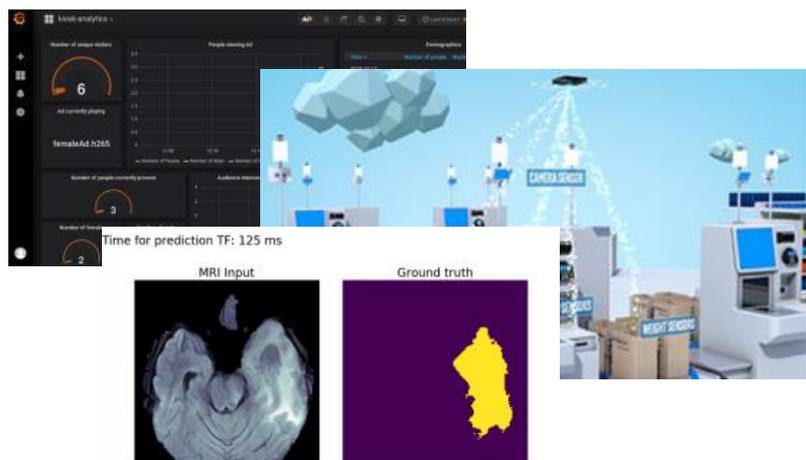
## PRE-TRAINED MODELS

[https://github.com/openvinotoolkit/open\\_model\\_zoo/tree/master/models](https://github.com/openvinotoolkit/open_model_zoo/tree/master/models)

# Demos Applications

Quickly get started with example demo applications

Take advantage of pre-built, open-sourced example implementations with step-by-step guidance and required components list



[3D Human Pose Estimation Python\\* Demo](#)  
[3D Segmentation Python\\* Demo](#)  
[Action Recognition Python\\* Demo](#)  
[BERT Question Answering Embedding Python\\* Demo](#)  
[BERT Question Answering Python\\* Demo](#)  
[Classification C++ Demo](#)  
[Colorization Demo](#)  
[Crossroad Camera C++ Demo](#)  
[Face Detection MTCNN Python\\* Demo](#)  
[Formula Recognition Python\\* Demo](#)  
[G-API Interactive Face Detection Demo](#)  
[Gaze Estimation Demo](#)  
[Gesture Recognition Python\\* Demo](#)  
[Handwritten Text Recognition Demo](#)  
[Human Pose Estimation C++ Demo](#)  
[Human Pose Estimation Python\\* Demo](#)  
[Image Deblurring Python\\* Demo](#)  
[Image Inpainting Python Demo](#)  
[Image Retrieval Python\\* Demo](#)  
[Image Segmentation C++ Demo](#)  
[Image Segmentation Python\\* Demo](#)  
[Image Translation Demo](#)  
[Instance Segmentation Python\\* Demo](#)  
[Interactive Face Detection C++ Demo](#)  
[Machine Translation Python\\* Demo](#)  
[MonoDepth Python Demo](#)

[Multi Camera Multi Target Python\\* Demo](#)  
[Multi-Channel Face Detection C++ Demo](#)  
[Multi-Channel Human Pose Estimation C++ Demo](#)  
[Multi-Channel Object Detection Yolov3 C++ Demo](#)  
[Object Detection C++ Demo](#)  
[Object Detection Python\\* Demo](#)  
[Pedestrian Tracker C++ Demo](#)  
[Place Recognition Python\\* Demo](#)  
[Security Barrier Camera C++ Demo](#)  
[Single Human Pose Estimation Demo \(top-down pipeline\)](#)  
[Smart Classroom C++ Demo](#)  
[Sound Classification Python\\* Demo](#)  
[Speech Recognition Demo](#)  
[Super Resolution C++ Demo](#)  
[TensorFlow\\* Object Detection Mask R-CNNs Segmentation C++ Demo](#)  
[Text Detection C++ Demo](#)  
[Text Spotting Python\\* Demo](#)  
[Text-to-speech Python\\* Demo](#)  
[Whiteboard Inpainting Demo](#)

## DEMO APPLICATIONS

[https://github.com/openvinotoolkit/open\\_model\\_zoo/tree/master/demos](https://github.com/openvinotoolkit/open_model_zoo/tree/master/demos)

# OpenVINO as execution provider

- You can use OpenVINO Inference Engine as backend of other DL Inference Frameworks such as Tensorflow or ONNX Runtime



- Benefit: the advantages of OpenVINO (multiple HW support and acceleration) in your favorite framework

# OpenVINO™ Integration with TensorFlow\*

Only 2 lines  
to be added  
to regular  
Tensorflow

```
1 # Installation steps
2 # more details : https://github.com/openvinotoolkit/openvino\_tensorflow
3 #pip3 install -U pip==21.0.1
4 #pip3 install -U tensorflow==2.4.1
5 #pip3 install openvino-tensorflow
6
7 # Import package and set backend
8 import openvino_tensorflow
9 openvino_tensorflow.set_backend('GPU')
10
11 # Load a TF Saved Model
12 model = tf.keras.models.load_model('resnet50_saved_model')
13
14 # Get the input size of the model
15 network_input_size = saved_model_loaded.input.shape()
16
17 # Resize the input image
18 resized_image = resize(input_image, network_input_size)
19
20 # Run inference
21 model.predict(resized_image)
```

CPU  
GPU  
MYRIAD  
VAD-M

# Pre-trained Models

Open-sourced repository of pre-trained models and support for public models



## Intel Pre-trained Models

[Action Recognition Models](#)  
[Classification Models](#)  
[Head Pose Estimation Models](#)  
[Human Pose Estimation Models](#)  
[Image Processing Models](#)

[Instance Segmentation Models](#)  
[Machine Translation Models](#)  
[Object Attribute Estimation Models](#)  
[Object Detection Models](#)  
[Optical Character Recognition](#)

[Models](#)  
[Question Answering Models](#)  
[Semantic Segmentation Models](#)  
[Text-to-speech Models](#)  
[Token Recognition Models](#)



## Public Pre-trained Models

[Action Recognition Models](#)  
[Classification Models](#)  
[Colorization Models](#)  
[Face Recognition Models](#)  
[Human Pose Estimation Models](#)  
[Image Inpainting Models](#)  
[Image Processing Models](#)  
[Image Translation Models](#)

[Instance Segmentation Models](#)  
[Monocular Depth Estimation Models](#)  
[Object Attribute Estimation Models](#)  
[Object Detection Models](#)  
[Optical Character Recognition Models](#)

[Place Recognition Models](#)  
[Semantic Segmentation Models](#)  
[Sound Classification Models](#)  
[Speech Recognition Models](#)  
[Style Transfer Models](#)  
[Text-to-speech Models](#)

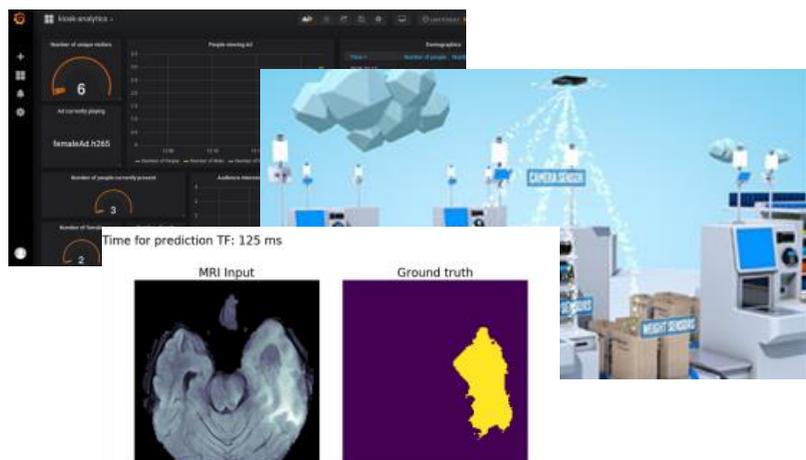
## PRE-TRAINED MODELS

[https://github.com/openvinotoolkit/open\\_model\\_zoo/tree/master/models](https://github.com/openvinotoolkit/open_model_zoo/tree/master/models)

# Demos Applications

Quickly get started with example demo applications

Take advantage of pre-built, open-sourced example implementations with step-by-step guidance and required components list



[3D Human Pose Estimation Python\\* Demo](#)  
[3D Segmentation Python\\* Demo](#)  
[Action Recognition Python\\* Demo](#)  
[BERT Question Answering Embedding Python\\* Demo](#)  
[BERT Question Answering Python\\* Demo](#)  
[Classification C++ Demo](#)  
[Colorization Demo](#)  
[Crossroad Camera C++ Demo](#)  
[Face Detection MTCNN Python\\* Demo](#)  
[Formula Recognition Python\\* Demo](#)  
[G-API Interactive Face Detection Demo](#)  
[Gaze Estimation Demo](#)  
[Gesture Recognition Python\\* Demo](#)  
[Handwritten Text Recognition Demo](#)  
[Human Pose Estimation C++ Demo](#)  
[Human Pose Estimation Python\\* Demo](#)  
[Image Deblurring Python\\* Demo](#)  
[Image Inpainting Python Demo](#)  
[Image Retrieval Python\\* Demo](#)  
[Image Segmentation C++ Demo](#)  
[Image Segmentation Python\\* Demo](#)  
[Image Translation Demo](#)  
[Instance Segmentation Python\\* Demo](#)  
[Interactive Face Detection C++ Demo](#)  
[Machine Translation Python\\* Demo](#)  
[MonoDepth Python Demo](#)

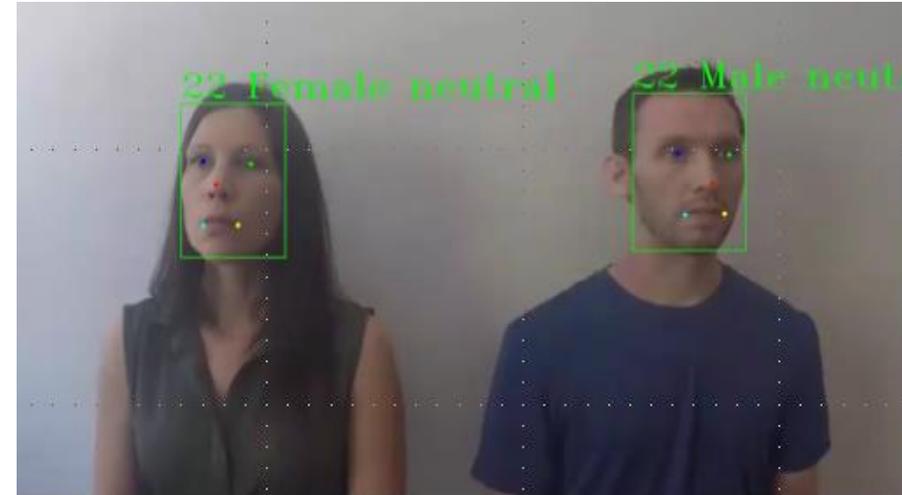
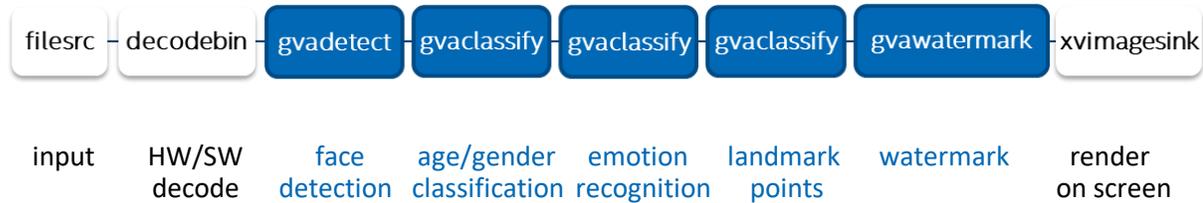
[Multi Camera Multi Target Python\\* Demo](#)  
[Multi-Channel Face Detection C++ Demo](#)  
[Multi-Channel Human Pose Estimation C++ Demo](#)  
[Multi-Channel Object Detection Yolov3 C++ Demo](#)  
[Object Detection C++ Demo](#)  
[Object Detection Python\\* Demo](#)  
[Pedestrian Tracker C++ Demo](#)  
[Place Recognition Python\\* Demo](#)  
[Security Barrier Camera C++ Demo](#)  
[Single Human Pose Estimation Demo \(top-down pipeline\)](#)  
[Smart Classroom C++ Demo](#)  
[Sound Classification Python\\* Demo](#)  
[Speech Recognition Demo](#)  
[Super Resolution C++ Demo](#)  
[TensorFlow\\* Object Detection Mask R-CNNs Segmentation C++ Demo](#)  
[Text Detection C++ Demo](#)  
[Text Spotting Python\\* Demo](#)  
[Text-to-speech Python\\* Demo](#)  
[Whiteboard Inpainting Demo](#)

## DEMO APPLICATIONS

[https://github.com/openvinotoolkit/open\\_model\\_zoo/tree/master/demos](https://github.com/openvinotoolkit/open_model_zoo/tree/master/demos)

# Video Analytics Pipeline

## Detection and Classification



```
gst-launch-1.0 filesrc location=/path/to/video.mp4 !
decodebin ! \
gvadetect model=face-detection-adas-0001.xml model-proc=face-detection-adas-0001.json ! queue ! \
gvaclassify device=CPU model=age-gender-recognition.xml model-proc=age-gender-recognition.json ! queue ! \
gvaclassify device=GPU model=emotions-recognition.xml model-proc=emotions-recognition.json ! queue ! \
gvaclassify device=MULTI:CPU,GPU model=landmarks-regression.xml model-proc=landmarks-regression.json ! queue ! \
gvawatermark ! ximagesink
```

# OpenVINO Notebooks

## Python Demos and Tutorials for OpenVINO

CPU

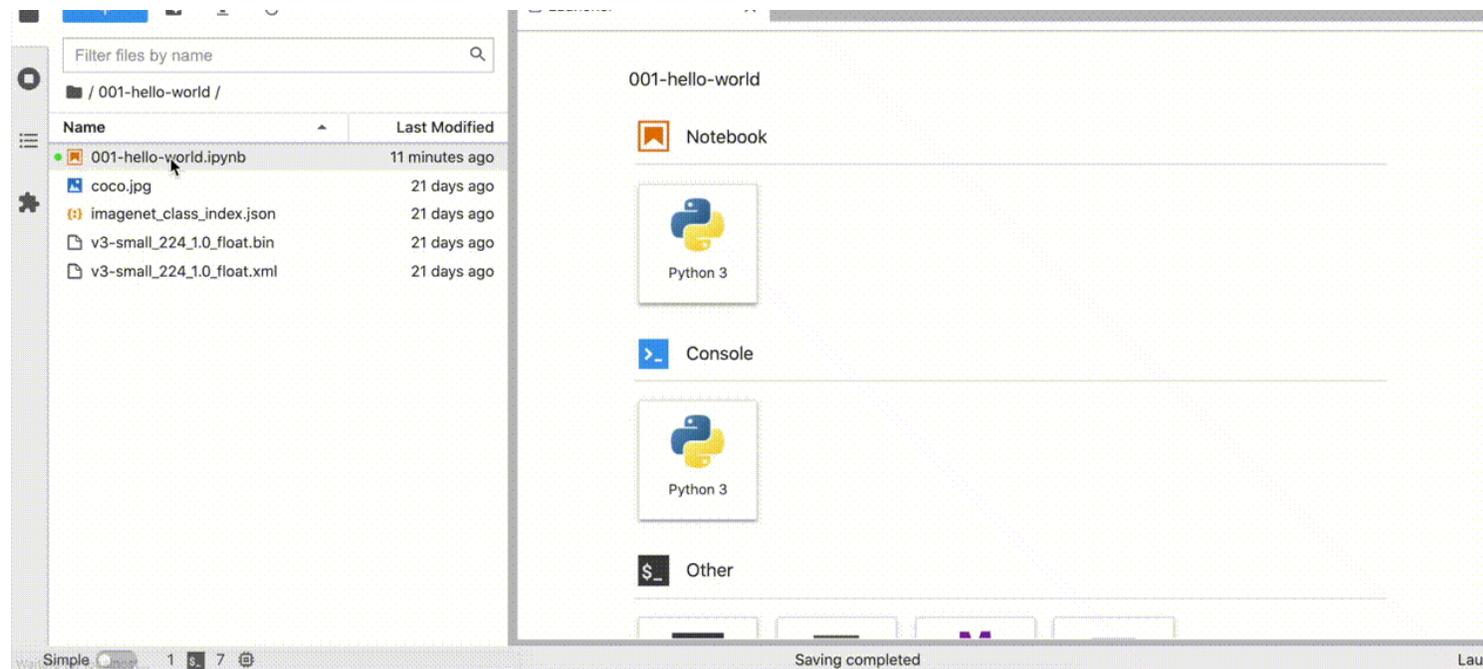
GPU

VPU

GNA

### How to get OpenVINO

### setup and running in 5 minutes?



OS  
SUPPORTED

WIN 10

WIN  
SERVER

LINUX

MAC

# Intel® DevCloud for the Edge

# Accelerate Test Cycles with the Intel® DevCloud for the Edge

A Development Sandbox for Developers, Researchers, and Startups to Test AI and Vision Workloads Remotely before Deployment.

## With the Intel® DevCloud for the Edge users can:

- **Prototype** on the latest hardware and software to future proof the solution
- **Benchmark** the customized AI application
- Run AI applications from **anywhere in the world**
- **Reduce** development time and cost

## [New] DL Workbench + Intel® DevCloud for the Edge

Developers can now graphically analyze models using the DL Workbench on Intel® DevCloud for the Edge (instead of local machine only) to compare, visualize and fine-tune a solution against multiple remote hardware configurations

For more information visit ► <https://devcloud.intel.com/edge/>

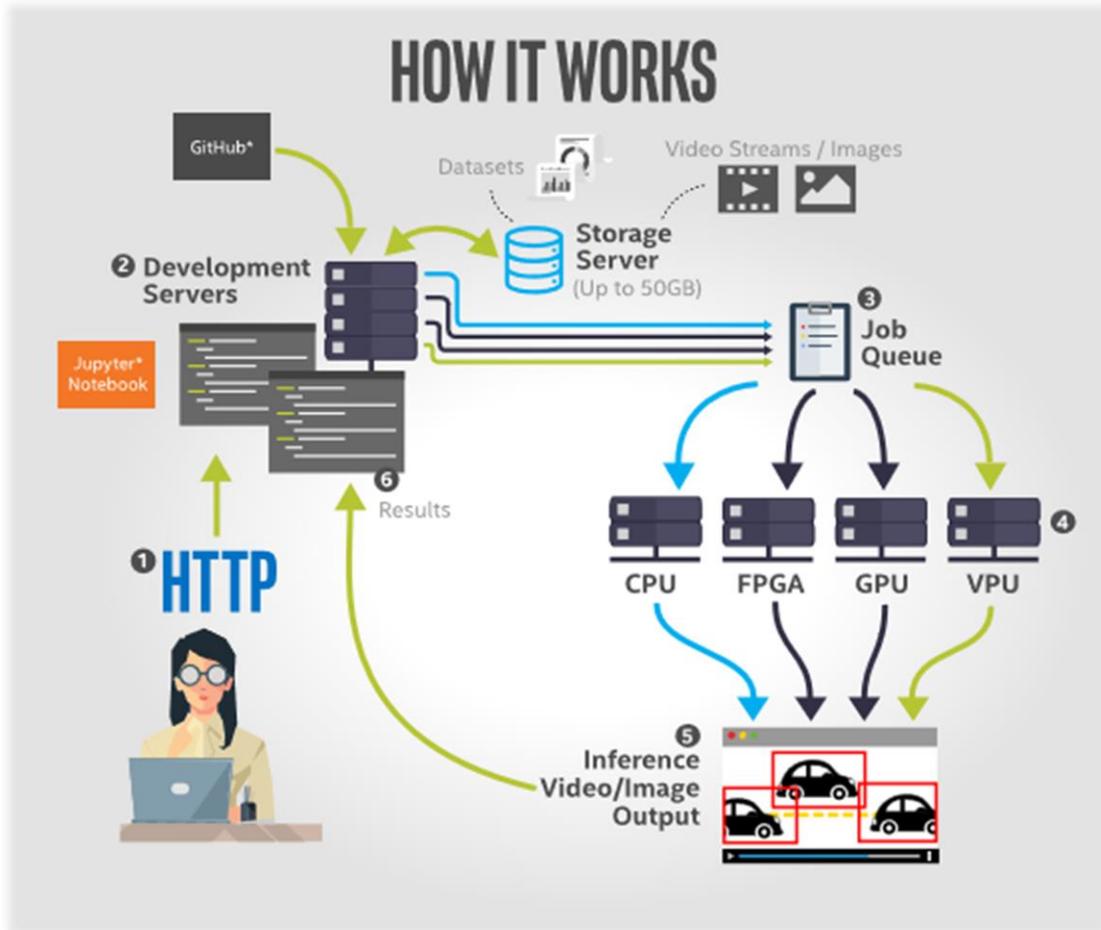


Deploy and scale



# Accelerate Time to Production with Intel® DevCloud for the Edge

See immediate AI Model performance across Intel's vast array of Edge Solutions



- **Instant, Global Access**  
Run AI applications from anywhere in the world
- **Prototype on the Latest Hardware and Software**  
Develop knowing you're using the latest Intel technology
- **Benchmark your Customized AI Application**  
Immediate feedback - frames per second, performance
- **Reduce Development Time and Cost**  
Quickly find the right compute for your edge solution

[Learn more](#)

[Sign up now for access](#)

# Demo

# Ready to get started?

Download directly from Intel for free

[Intel® Distribution of OpenVINO™ toolkit](#)  
(Recommended)

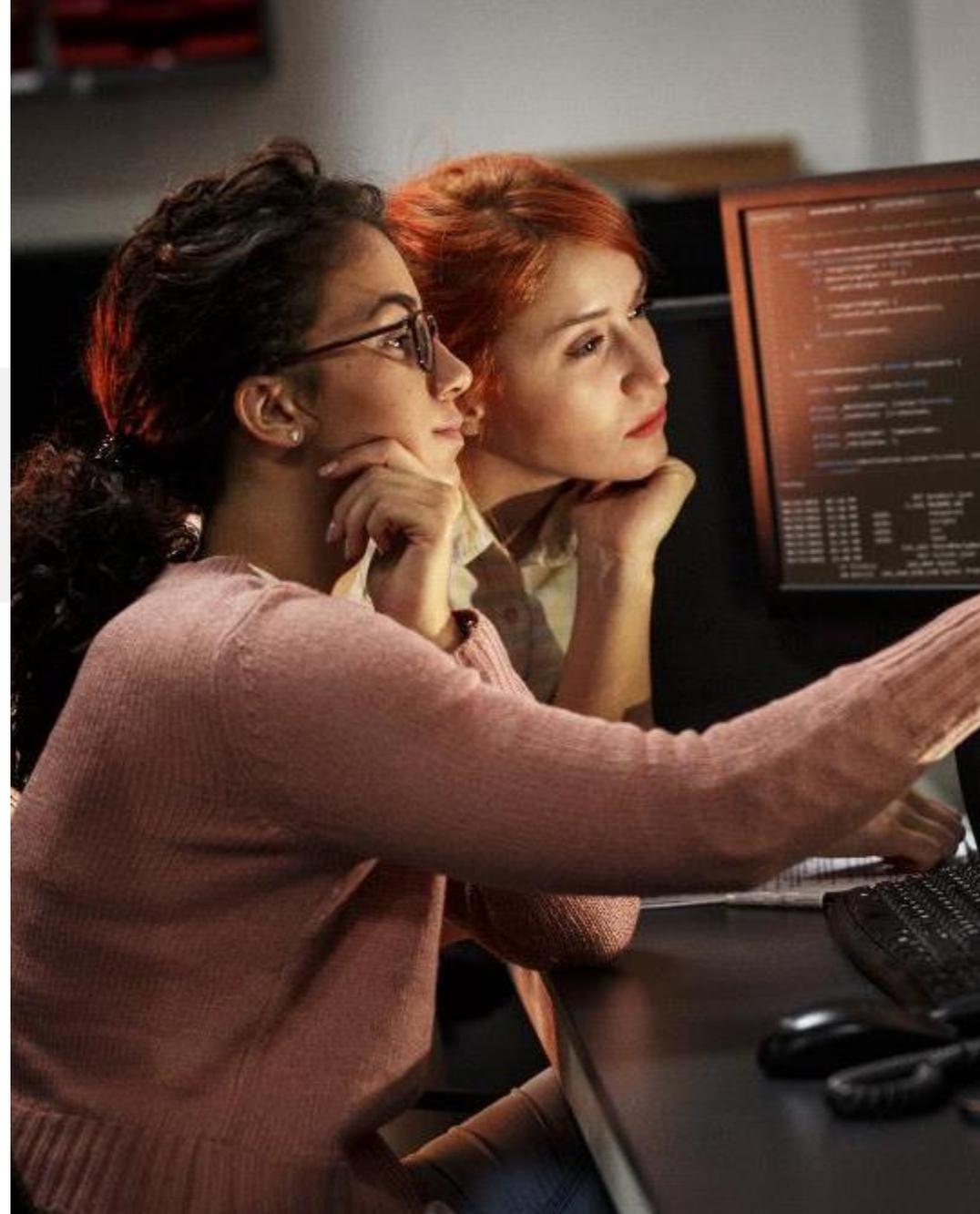
*Also available from*

Intel's Edge Software Hub | Intel® DevCloud for the Edge | PIP |  
DockerHub | Dockerfile | Anaconda Cloud | YUM | APT

*Build from source*

GitHub | Gitee (for China)

[Choose & Download](#)



# Questions?

intel®