

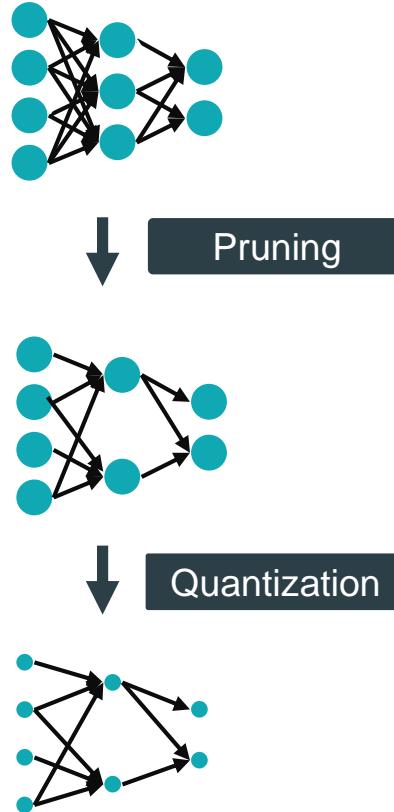
# Xilinx Edge AI Solution

Andy Luo, AI/ML Product Marketing  
[andy.luo@xilinx.com](mailto:andy.luo@xilinx.com)

Jan 2019



# Unique, Patented Deep Learning Acceleration Techniques



> Best paper awards for breakthrough DL acceleration

> Xilinx's compression technology

>> Reduce DL accelerator footprint into smaller devices

>> Increase performance per watt (higher performance and/or lower energy)



Unique Pruning Technology Provides a Significant Competitive Advantage

# Xilinx Solution Stack for Edge/Embedded ML

## Models



## Framework

Caffe

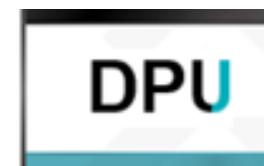


Darknet



TensorFlow

## Tools & IP



## HW Platforms



Z7020 Board



Z7020 SOM



ZU2 SOM



ZU2/3 Card



ZU9 Card



ZCU102



ZCU104



Ultra96

# DNNDK – Deep Neural Network Development Kit

- > DECENT (DEep ComprEssioN Tool)
- > DNNC (Deep Neural Network Compiler)
- > Runtime N<sup>2</sup>Cube (Cube of Neural Network)
- > Profiler DSight



# Framework Support

## Caffe

- Pruning
- Quantization
- Compilation

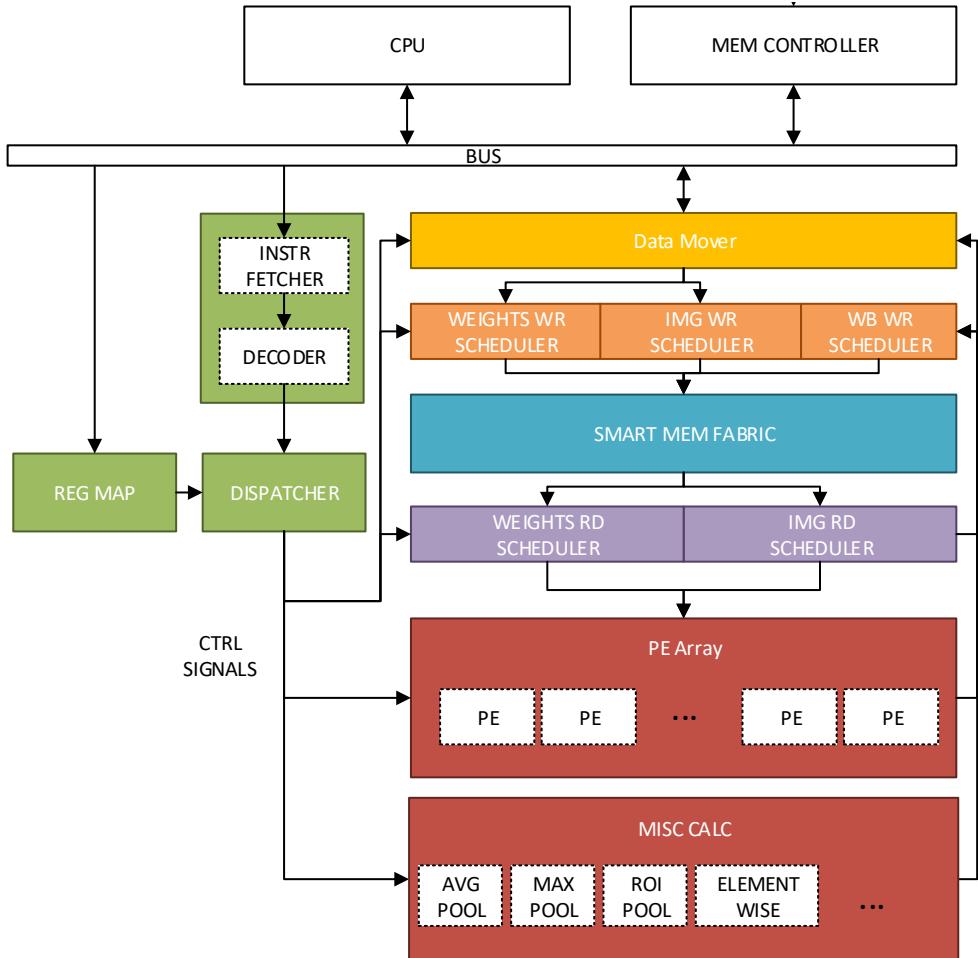


- Pruning
- Quantization
- Convertor to Caffe

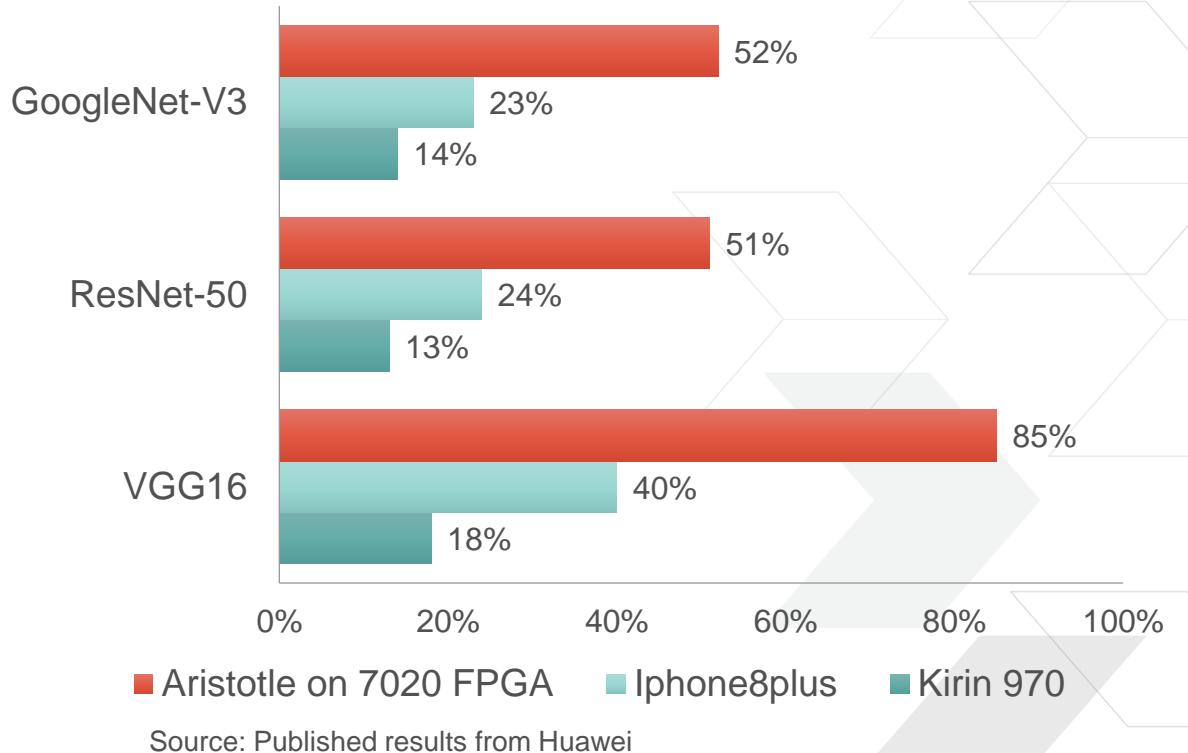


- Quantization & Compilation
  - Beta version
- Pruning
  - Beta version

# DPU IP with High Efficiency



Utilization > 50% for mainstream neural networks



# Supported Operators

- Conv
  - Dilation
- Pooling
  - Max
  - Average
- ReLU / Leaky Relu/ Relu6
- Full Connected (FC)
- Batch Normalization
- Concat
- Elementwise
- Deconv
- Depthwise conv
- Mean scale
- Upsampling
- Split
- Reorg
- Resize (Optional)
- Softmax (Optional)
- Sigmoid (Optional)

# Constraints Between Layers

Next Layer Type \ Current Layer Type	Conv	Deconv	Depth-wise Conv	Inner Product	Max Pooling	Ave Pooling	BN	ReLU	LeakyReLU	Element-wise	Concat	As Input	As Output
Conv	●	●	○	●	●	○	●	●	○	●	●	●	●
Deconv	●	●	○	●	●	○	●	●	○	●	●	●	●
Depth-wise Conv	●	●	○	●	●	○	●	●	○	●	●	●	●
Inner Product	●	●	○	●	●	○	●	●	○	●	●	●	●
Max Pooling	●	●	○	●	●	○	○	×	×	●	●	●	●
Ave Pooling	○	○	○	○	○	○	○	×	×	○	○	○	○
BN	●	●	○	●	●	○	○	●	×	●	●	○	○
ReLU	●	●	○	●	●	○	○	×	×	●	●	--	●
LeakyReLU	○	○	○	○	○	○	○	×	×	○	○	--	○
Element-wise	●	●	○	●	●	○	○	●	○	●	●	--	●
Concat	●	●	○	●	●	○	○	×	×	●	●	--	●

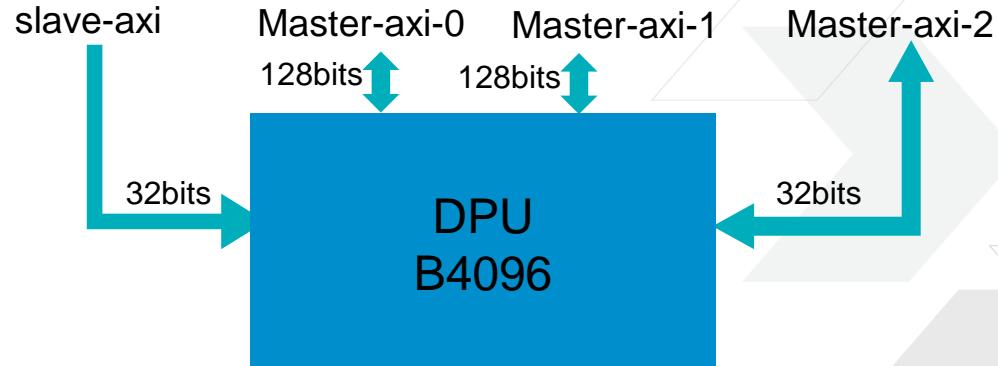
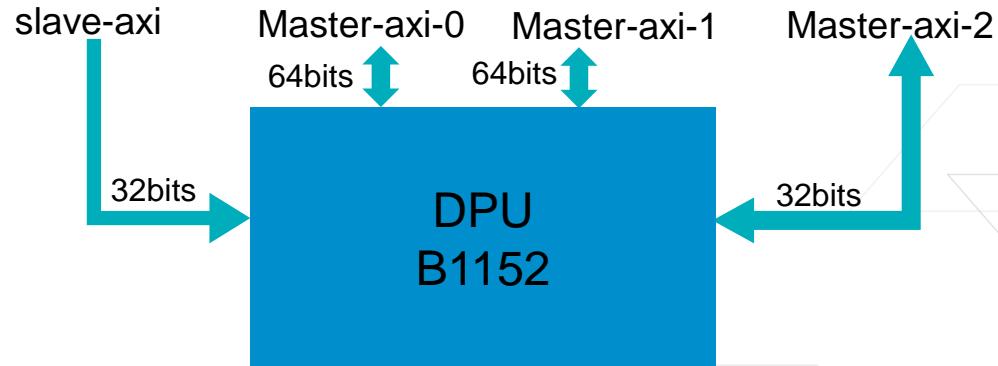
● : Support

✗: Not support

○ : Support when selecting additional features

# DPU Typical Options & Interfaces

- > **3-level parallelism is exploited**
  - >> Pixel \* input channel \* output channel
- > **Small core - B1152**
  - >> Parallelism: 4\*12\*12
  - >> target Z7020/ZU2/ZU3
- > **Big core - B4096**
  - >> Parallelism: 8\*16\*16
  - >> Target ZU5 and above



# DPU vs DPU\_EU

## ➤ DPU

- Just include one clock domain
- Instructions: Convolution, Deconvolution, Depthwise Convolution, MaxPool, AveragePool, Elementwise, Softmax, Sigmoid.....

## ➤ DPU\_EU/DPU\_EU\_LP

- Include two clock domains
- Use DSP DDR technique
- Adopt cascade technology to reduce resources
- Use gated clock to reduce power consumption

\*DPU\_EU\_LP in development

# DPU\_EU Utilization

More DSP

Arch	LUTs	Registers	BRAM*	DSP
B512	17951	28280	69.5	97
B800	20617	35065	87	141
B1024	22327	39000	101.5	193
B1152	22796	40276	117.5	193
B1600	26270	50005	123	281
B2304	29592	57549	161.5	385
B3136	33266	69110	203.5	505
B4096	37495	84157	249.5	641

More LUT

Arch	LUTs	Registers	BRAM*	DSP
B512	20759	33572	69.5	66
B1024	29155	49823	101.5	130
B1152	30043	49588	117.5	146
B1600	33130	60739	123	202
B2304	37055	72850	161.5	290
B3136	41714	86132	203.5	394
B4096	44583	99791	249.5	514

DPU provides flexible option depending on costumer's resources and continues to improve

\* URAM also can be used by DPU if device supports, every URAM is roughly used as 3.7 BRAM

# DPU\_EU Utilization

*LeakyRelu not enabled*

Arch	LUTs	Registers	BRAM*	DSP
B512	17951	28280	69.5	97
B800	20617	35065	87	141
B1024	22327	39000	101.5	193
B1152	22796	40276	117.5	193
B1600	26270	50005	123	281
B2304	29592	57549	161.5	385
B3136	33266	69110	203.5	505
B4096	37495	84157	249.5	641

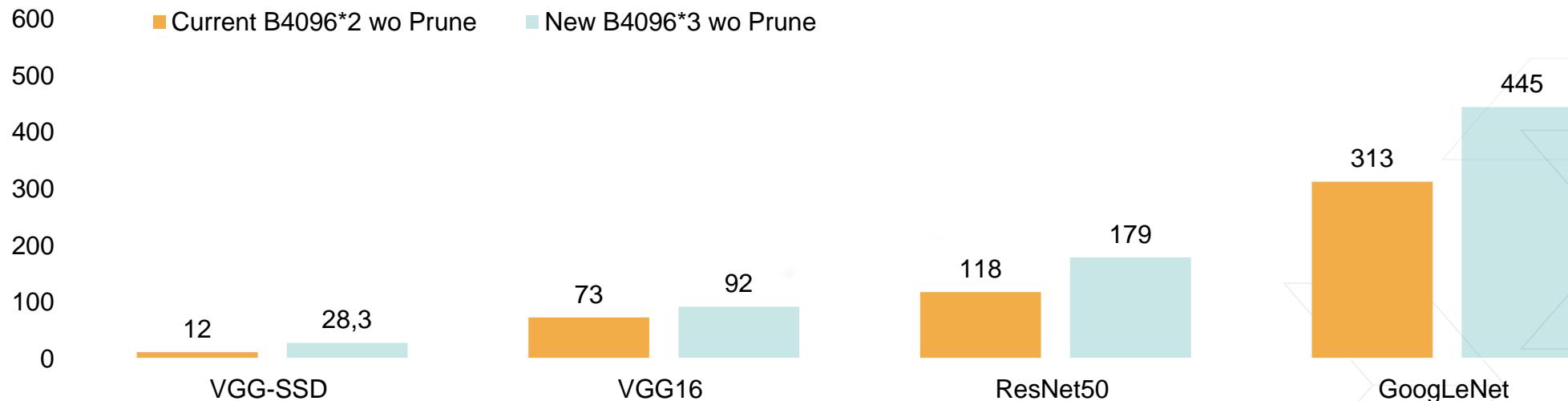
*LeakyRelu enabled*

Arch	LUTs	Registers	BRAM*	DSP
B512	18371	28292	69.5	97
B800	21162	35079	87	141
B1024	22759	39012	101.5	193
B1152	23453	40292	117.5	193
B1600	26817	50019	123	281
B2304	30268	57565	161.5	385
B3136	34032	69125	203.5	505
B4096	38392	84173	249.5	641

\* URAM also can be used by DPU if device supports, every URAM is roughly used as 3.7 BRAM

# Perf Improvement with DPU\_EU

## Performance Comparison (FPS)



\*The FPS of VGG-SSD of end to end performance

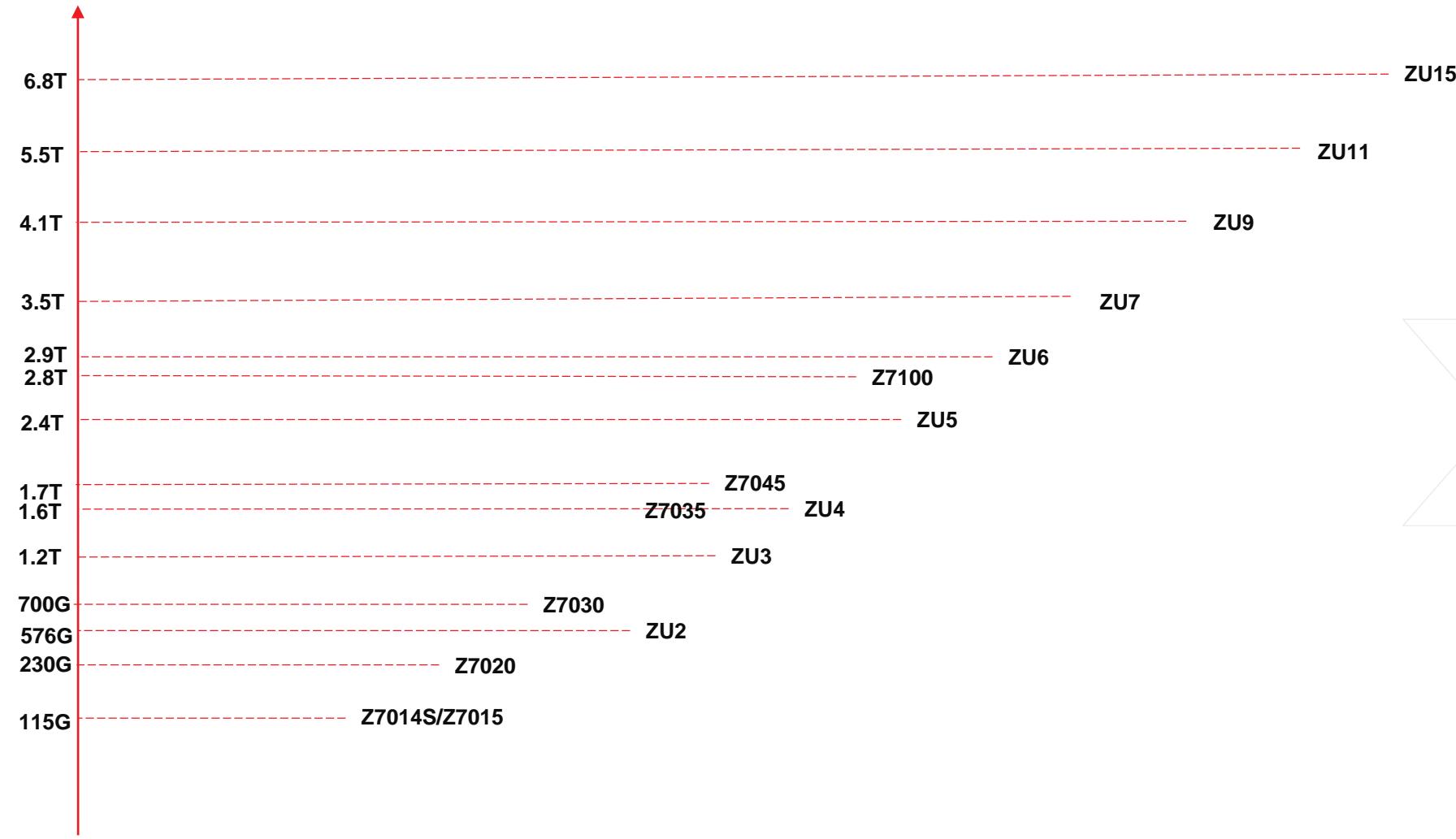
\*The FPS of VGG16/ResNet50/GoogLeNet is of CONV part (w/o FC layer)

## Resource Utilization Comparison

	DSP	LUT	FF	BRAM
DPU B4096*2	2048	156744	224650	501
DPU_EU B4096*3	1926	110311	255020	748.5

# DPU Scalability

Peak INT8 OPS\*



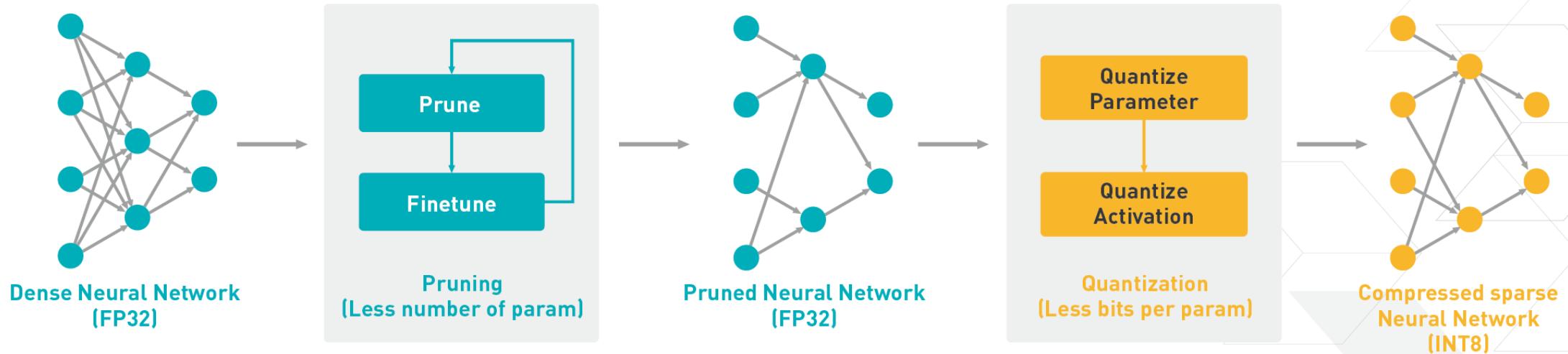
\* With heterogenous DPUs

# DNNDK Dev Flow

Five Steps  
with DNNDK

- 01 Model Compression
- 02 Model Compilation
- 03 Programming
- 04 Hybrid Compilation
- 05 Execution

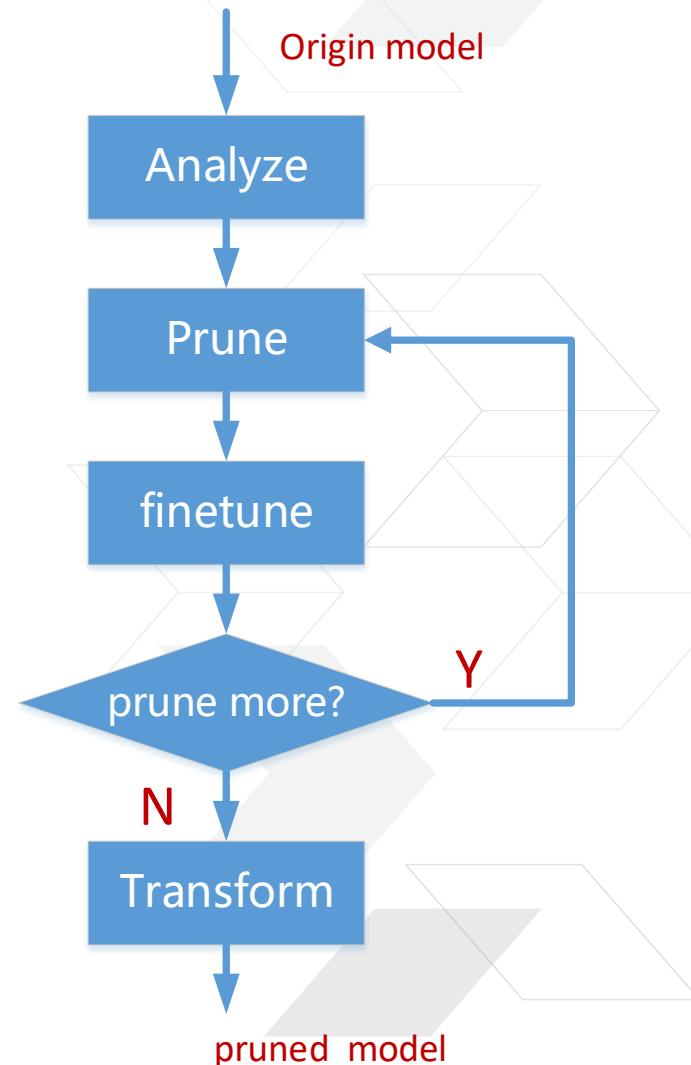
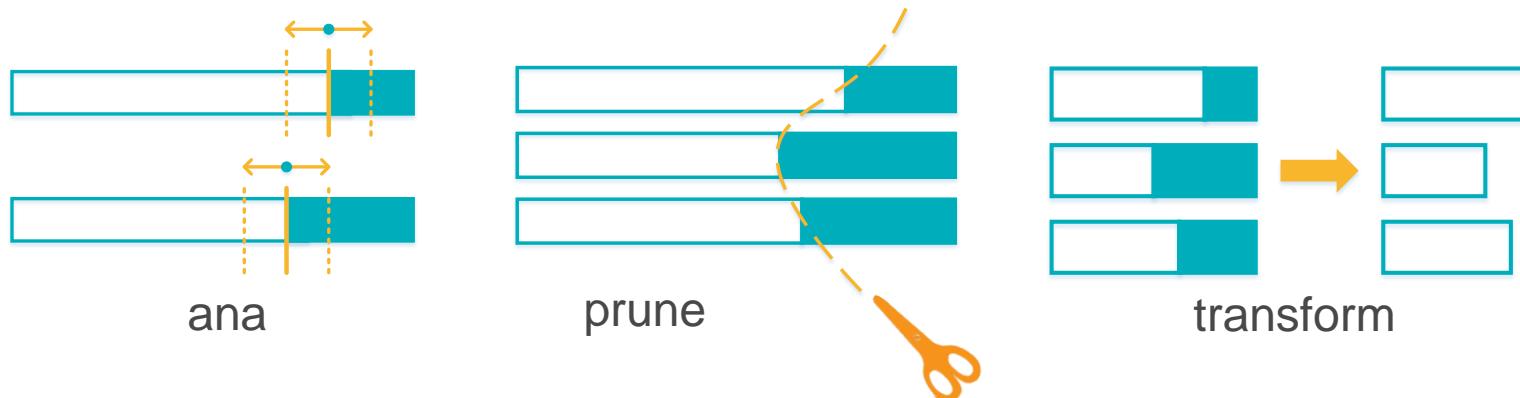
# DECENT – Xilinx Deep Compression Tool



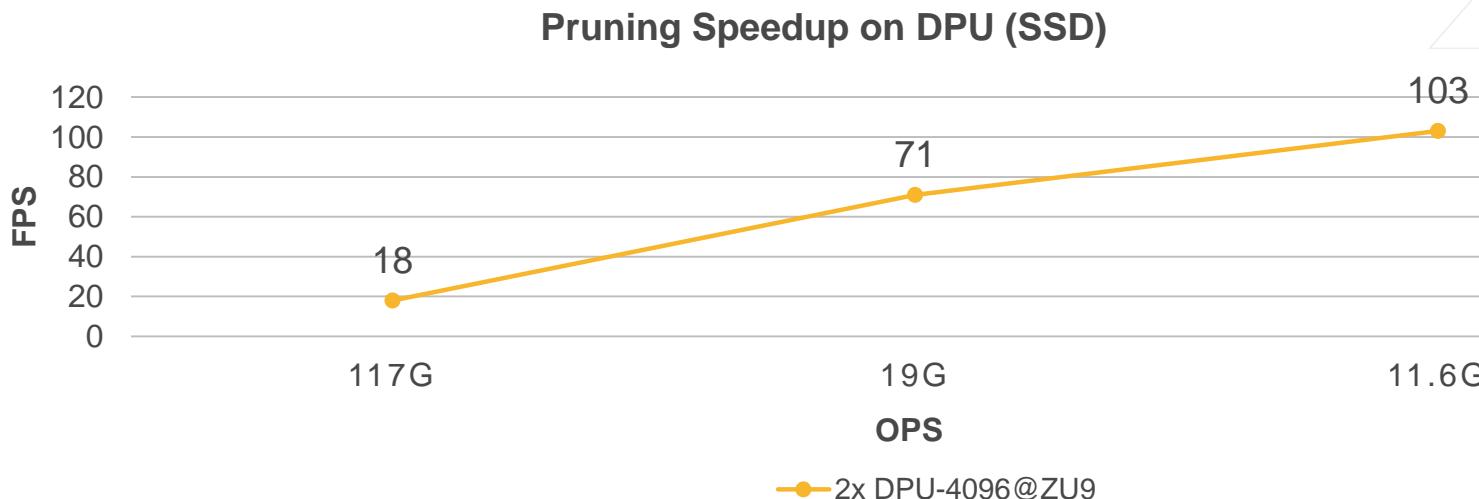
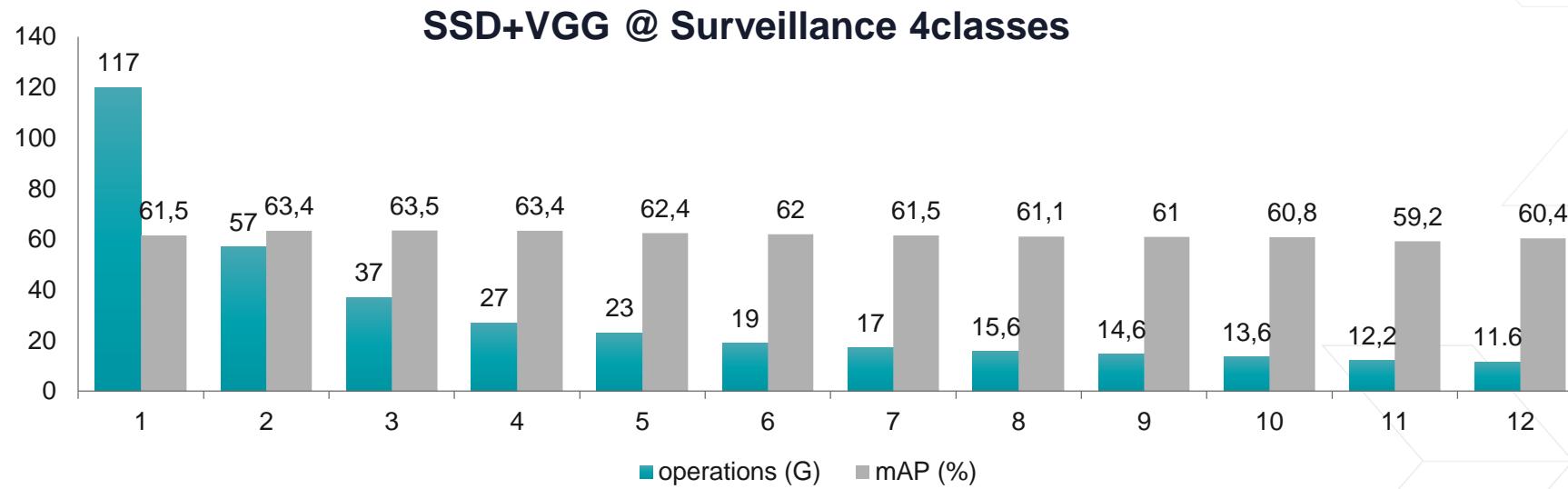
# Pruning Tool – decent\_p

## > 4 commands in decent\_p

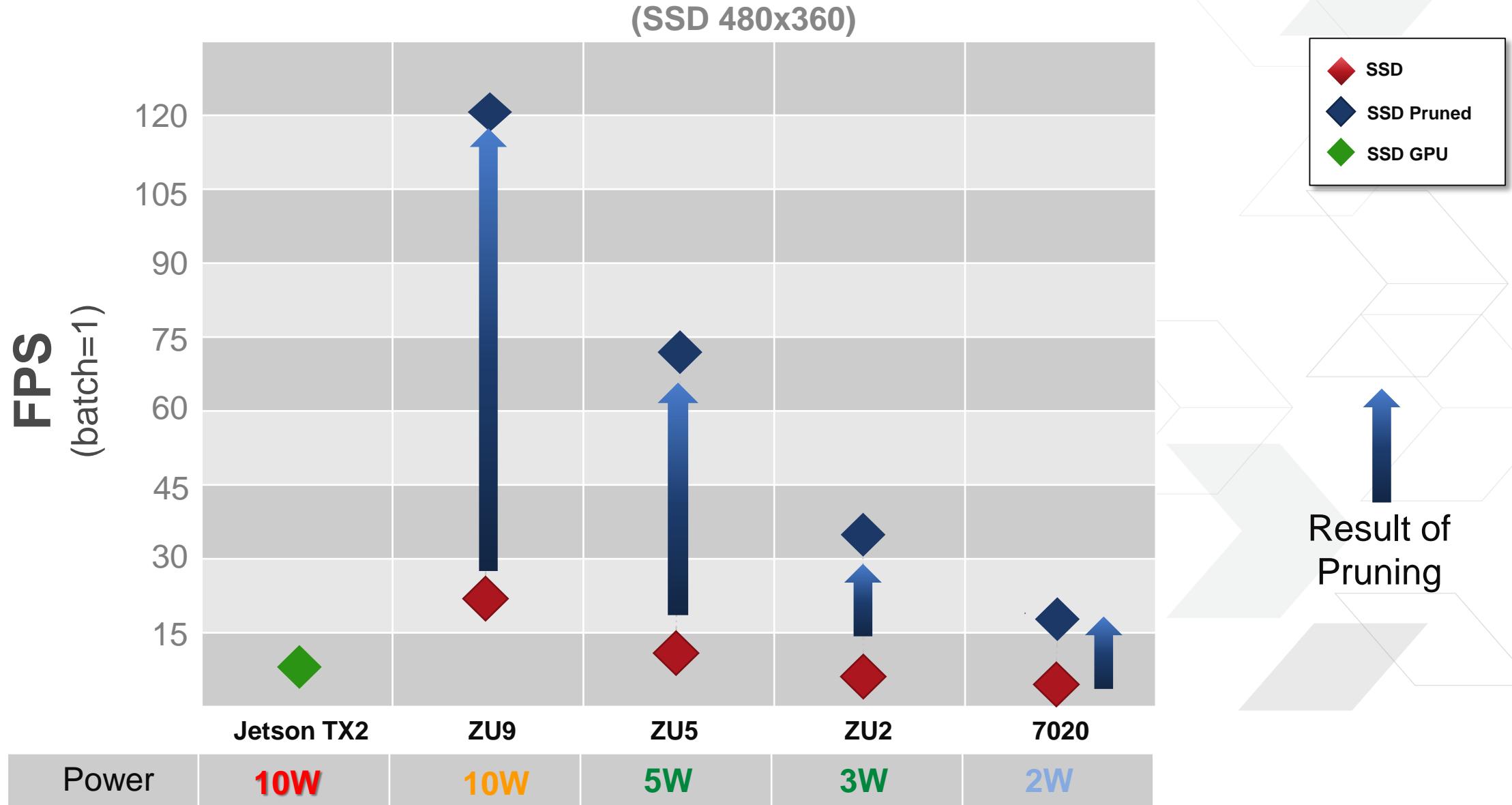
- >> Ana
  - analyze the network
- >> Prune
  - prune the network according to config
- >> Finetune
  - finetune the network to recover accuracy
- >> Transform
  - transform the pruned model to regular model



# Pruning Example - SSD



# Pruning Makes Big Difference



# Pruning Results

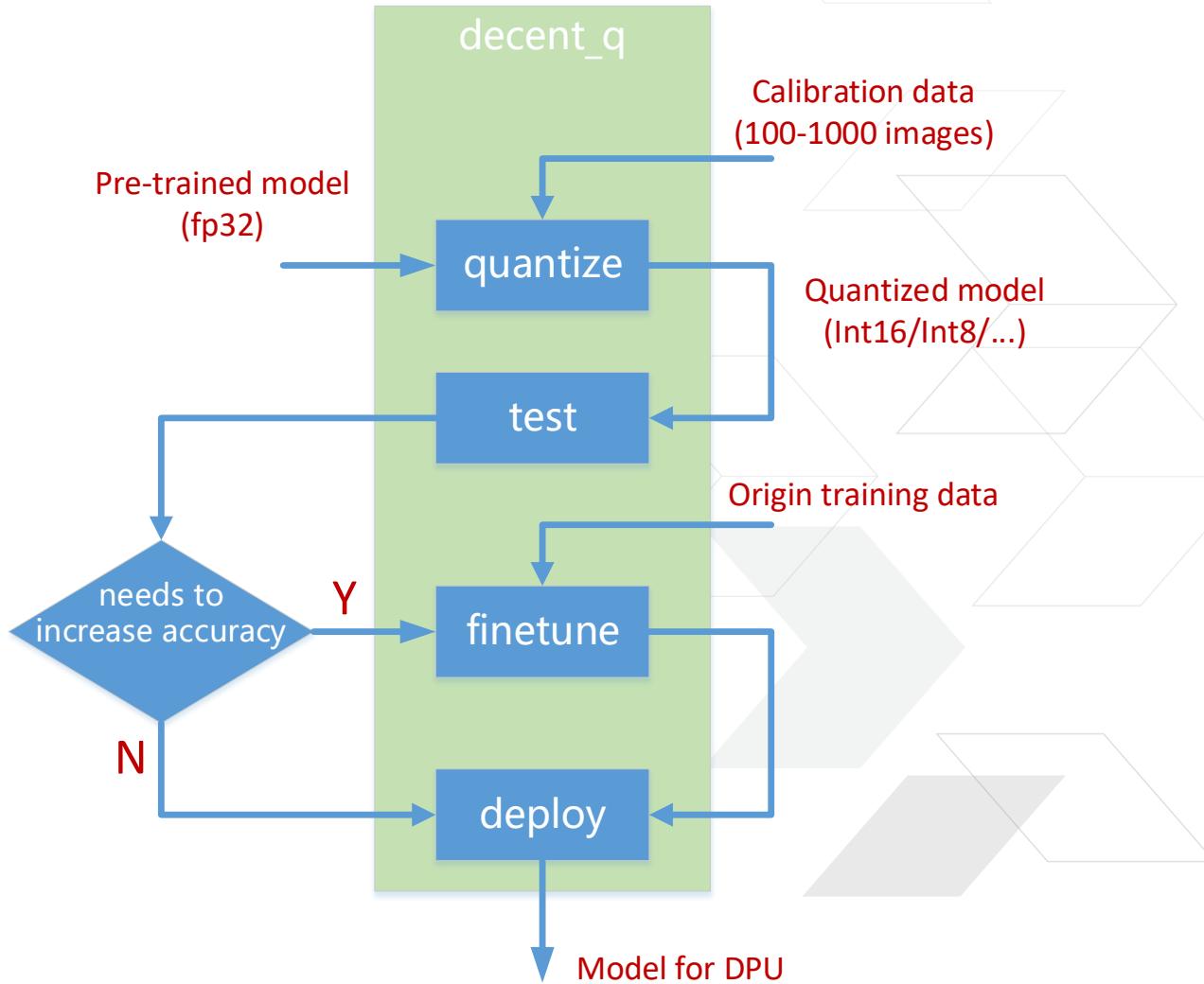
Classification Networks	Baseline	Pruning Result 1			Pruning Result 2		
	Top-5	Top-5	ΔTop5	ratio	Top-5	ΔTop5	ratio
Resnet50 [7.7G]	91.65%	91.23%	-0.42%	40%	90.79%	-0.86%	32%
Inception_v2 [4.0G]	91.07%	90.37%	-0.70%	60%	90.07%	-1.00%	55%
SqueezeNet [778M]	83.19%	82.46%	-0.73%	89%	81.57%	-1.62%	75%

Detection Networks	Baseline mAP	Pruning Result 1			Pruning Result 2		
		mAP	ΔmAP	ratio	mAP	ΔmAP	ratio
DetectNet [17.5G]	44.46	45.7	+1.24	63%	45.12	+0.66	50%
SSD+VGG [ 117G]	61.5	62.0	+0.5	16%	60.4	-1.1	10%
[A] SSD+VGG [ 173G]	57.1	58.7	+1.6	40%	56.6	-0.5	12%
[B] Yolov2 [ 198G]	80.4	81.9	+1.5	28%	79.2	-1.2	7%

# Quantization Tool – decent\_q

- > 4 commands in decent\_q
  - >> quantize
    - Quantize network
  - >> test
    - Test network accuracy
  - >> finetune
    - Finetune quantized network
  - >> deploy
    - Generate model for DPU
- > Data
  - >> Calibration data
    - Quantize activation
  - >> Training data
    - Further increase accuracy



# Quantization Results

## > Uniform Quantization

- >> 8-bit for both weights and activation
- >> A small set of images for calibration

Networks	Float32 baseline		8-bit Quantization			
	Top1	Top5	Top1	ΔTop1	Top5	ΔTop5
Inception_v1	66.90%	87.68%	66.62%	-0.28%	87.58%	-0.10%
Inception_v2	72.78%	91.04%	72.40%	-0.38%	90.82%	-0.23%
Inception_v3	77.01%	93.29%	76.56%	-0.45%	93.00%	-0.29%
Inception_v4	79.74%	94.80%	79.42%	-0.32%	94.64%	-0.16%
ResNet-50	74.76%	92.09%	74.59%	-0.17%	91.95%	-0.14%
VGG16	70.97%	89.85%	70.77%	-0.20%	89.76%	-0.09%
Inception-ResNet-v2	79.95%	95.13%	79.45%	-0.51%	94.97%	-0.16%

# DNNDK API

dpuOpen()  
dpuClose()  
dpuLoadKernel()  
dpuDestroyKernel()  
dpuCreateTask()  
dpuRunTask()  
dpuDestroyTask()  
dpuEnableTaskProfile()  
dpuGetTaskProfile()  
dpuGetNodeProfile()  
dpuGetInputTensor()  
dpuGetInputTensorAddress()  
dpuGetInputTensorSize()  
dpuGetInputTensorScale()  
dpuGetInputTensorHeight()  
dpuGetInputTensorWidth()  
dpuGetInputTensorChannel()  
dpuGetOutputTensor()  
dpuGetOutputTensorAddress()

dpuGetOutputTensorSize()  
dpuGetOutputTensorScale()  
dpuGetOutputTensorHeight()  
dpuGetOutputTensorWidth()  
dpuGetOutputTensorChannel()  
dpuGetTensorSize()  
dpuGetTensorAddress()  
dpuGetTensorScale()  
dpuGetTensorHeight()  
dpuGetTensorWidth()  
dpuGetTensorChannel()  
dpuSetInputTensorInCHWInt8()  
dpuSetInputTensorInCHWF32()  
dpuSetInputTensorInHWCI8()  
dpuSetInputTensorInHWCFP32()  
dpuGetOutputTensorInCHWInt8()  
dpuGetOutputTensorInCHWF32()  
dpuGetOutputTensorInHWCI8()  
dpuGetOutputTensorInHWCFP32()

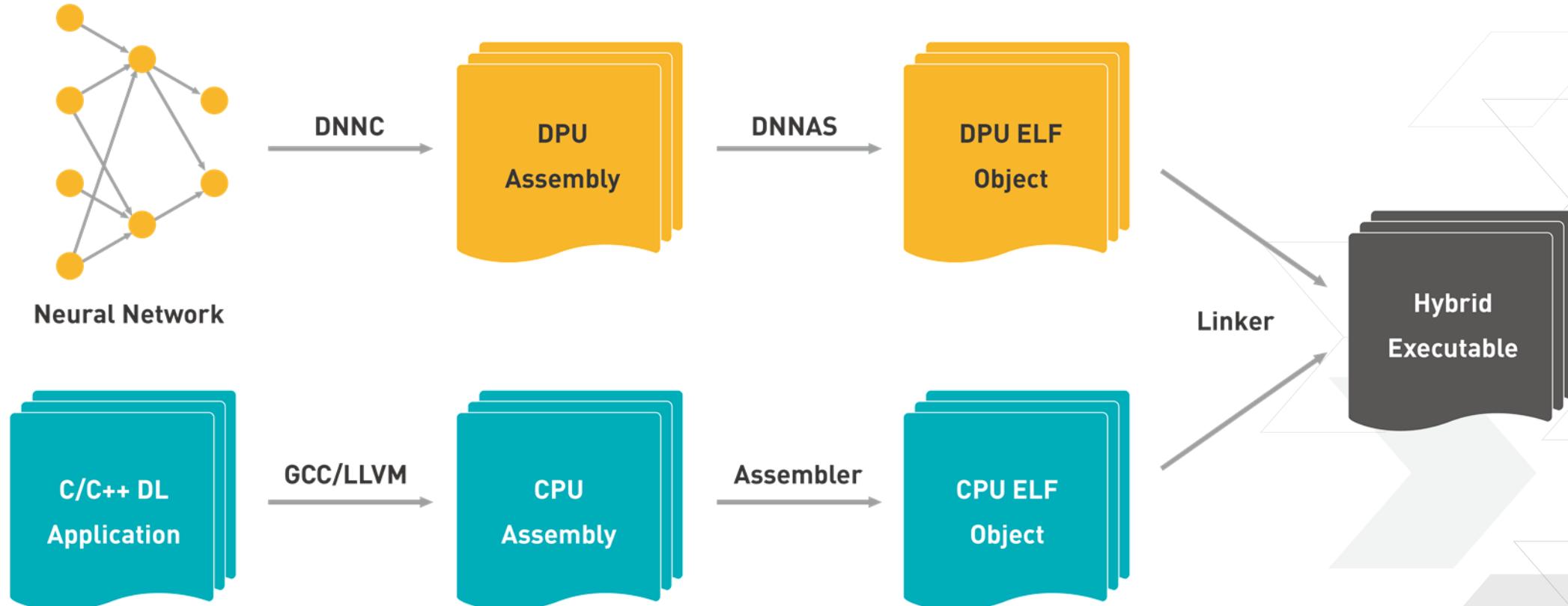
> **High-level Tensor-based APIs**

> **Please refer to DNNDK User Guide**

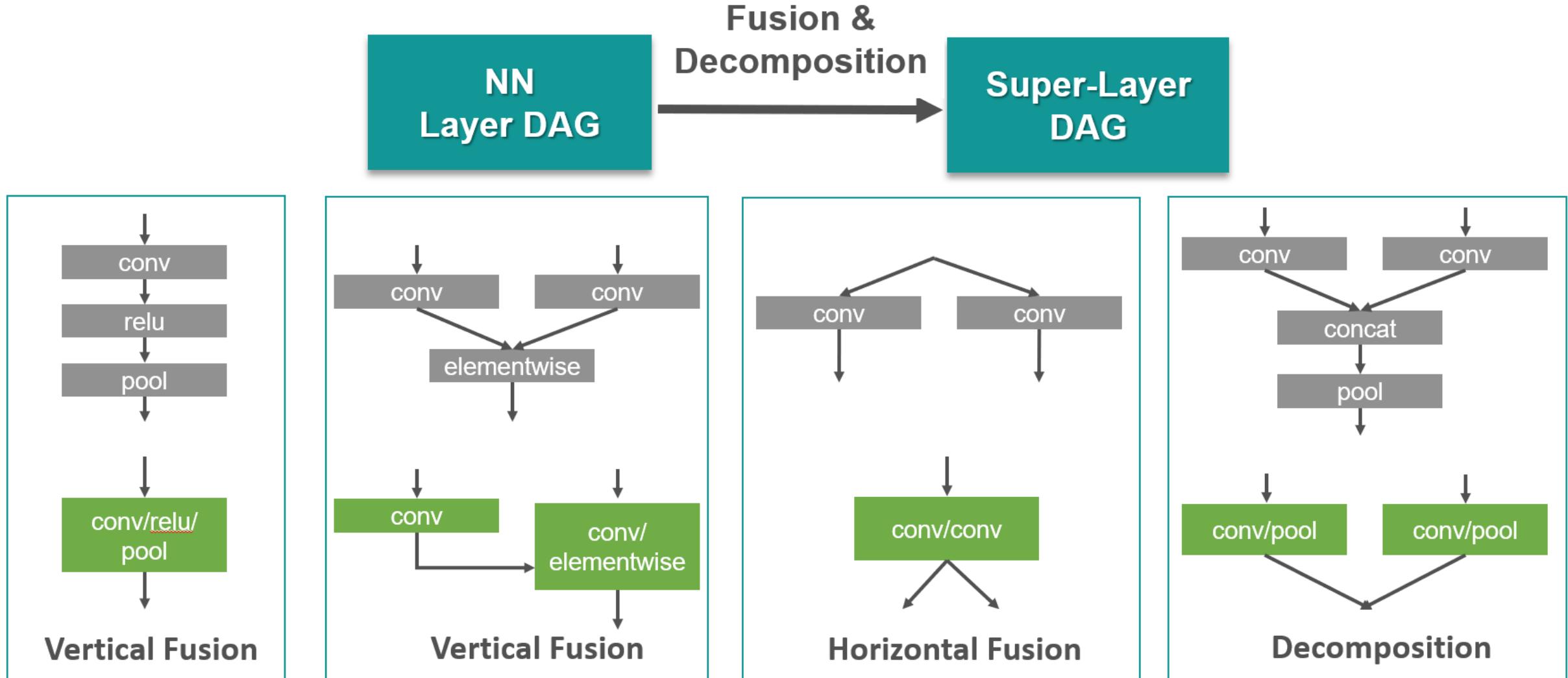
# Programming with DNNDK API

```
1 int main(int argc, char *argv[])
2 {
3     DPUKernel *kernel_conv;
4     DPUKernel *kernel_fc;
5     DPUTask *task_conv;
6     DPUTask *task_fc;
7     char *input_addr;
8     char *output_addr;
9
10    /* DNNDK API to attach to DPU driver */
11    dpuInit();
12
13    /* DNNDK API to create DPU kernels for CONV & FC networks */
14    kernel_conv = dpuLoadKernel("resnet50_conv", 224, 224);
15    kernel_fc = dpuLoadKernel("resnet50_fc", 1, 1);
16
17    /* Create tasks from CONV & FC kernels */
18    task_conv = dpuCreateTask(kernel_conv);
19    task_fc = dpuCreateTask(kernel_fc);
20
21    /* Set input tensor for CONV task and run */
22    input_addr = dpuGetTensorAddress(dpuGetTaskInputTensor(task_conv));
23    setInputImage(Mat &image, input_addr);
24    dpuRunTask(task_conv);
25    output_addr = dpuGetTensorAddress(dpuGetTaskOutputTensor(task_conv));
26
27    /* Run average pooling layer on CPU */
28    run_average_pooling(output_addr);
29
30    /* Set input tensor for FC task and run */
31    input_addr = dpuGetTensorAddress(dpuGetTaskInputTensor(task_fc));
32    setFCInputData(task_fc, input_addr);
33    dpuRunTask(task_fc);
34    output_addr = dpuGetTensorAddress(dpuGetTaskOutputTensor(task_fc));
35
36    /* Diaplay the Classification result from FC task */
37    displayClassificationResult(output_addr);
38
39    /* DNNDK API to destroy DPU tasks/kernels */
40    dpuDestroyTask(task_conv);
41    dpuDestroyTask(task_fc);
42
43    dpuDestroyKernel(kernel_conv);
44    dpuDestroyKernel(kernel_fc);
45
46    /* DNNDK API to dettach from DPU driver and free DPU resources */
47    dpuFini();
48
49    return 0;
50 }
```

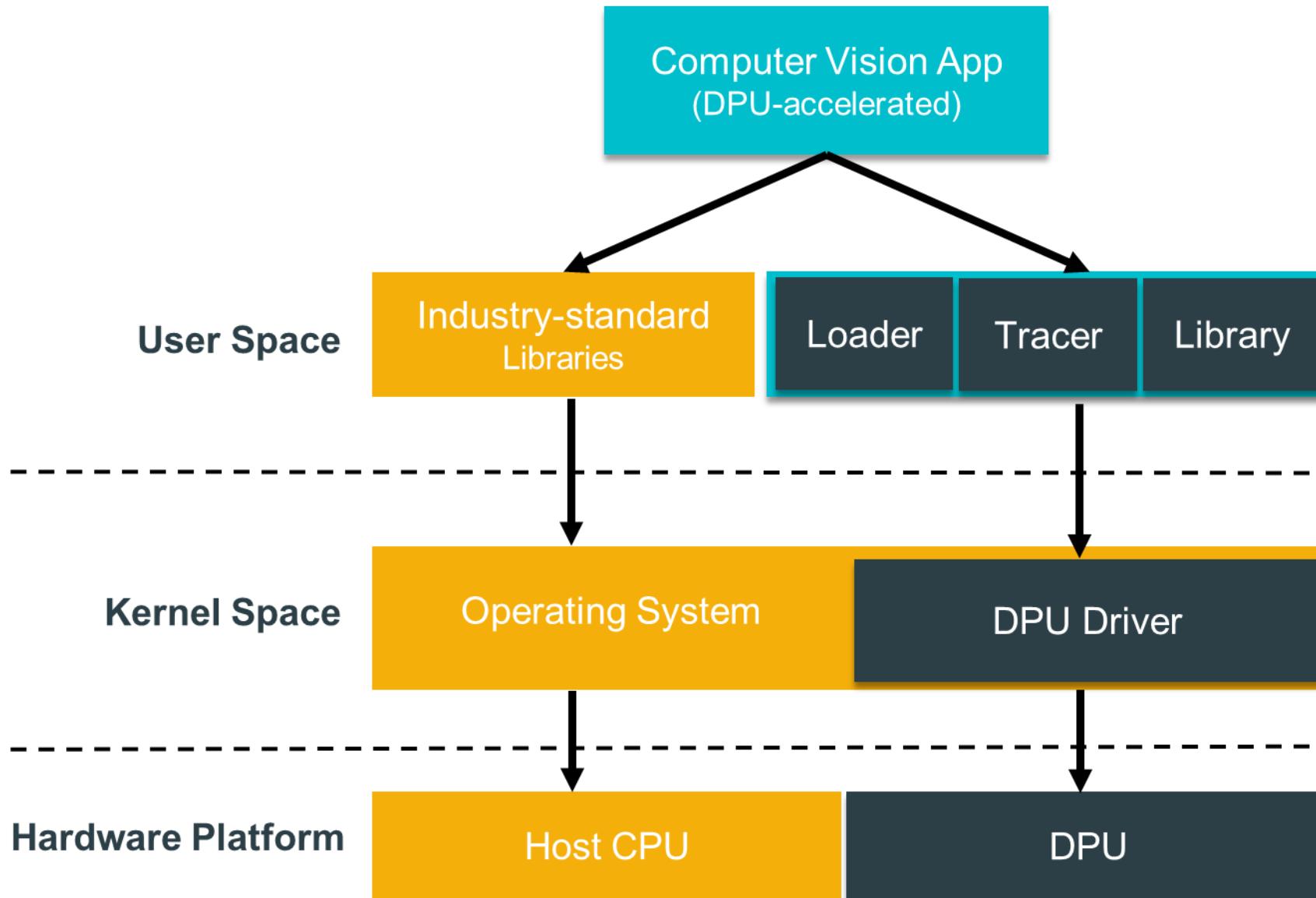
# DNNDK Hybrid Compilation Model



# Optimization in DNNC



# DNNDK Runtime Engine



## Runtime N<sup>2</sup>Cube

- Library
- Loader
- Tracer
- Driver

# Supported Networks

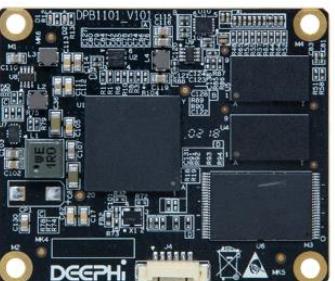
Application	Module	Algorithm	Model Development	Compression	Deployment
Face	Face detection	SSD, Densebox	✓	✓	✓
	Landmark Localization	Coordinates Regression	✓	N / A	✓
	Face recognition	ResNet + Triplet / A-softmax Loss	✓	✓	✓
	Face attributes recognition	Classification and regression	✓	N / A	✓
Pedestrian	Pedestrian Detection	SSD	✓	✓	✓
	Pose Estimation	Coordinates Regression	✓	✓	✓
	Person Re-identification	ResNet + Loss Fusion	✓		
Video Analytics	Object detection	SSD, RefineDet	✓	✓	✓
	Pedestrian Attributes Recognition	GoogleNet	✓	✓	✓
	Car Attributes Recognition	GoogleNet	✓	✓	✓
	Car Logo Detection	DenseBox	✓	✓	
	Car Logo Recognition	GoogleNet + Loss Fusion	✓	✓	
	License Plate Detection	Modified DenseBox	✓	✓	✓
	License Plate Recognition	GoogleNet + Multi-task Learning	✓	✓	✓
ADAS/AD	Object Detection	SSD, YOLOv2, YOLOv3	✓	✓	✓
	3D Car Detection	F-PointNet, AVOD-FPN	✓		
	Lane Detection	VPGNet	✓	✓	✓
	Traffic Sign Detection	Modified SSD	✓		
	Semantic Segmentation	FPN	✓	✓	✓
	Driveable Space Detection	MobilenetV2-FPN	✓		
	Multi-task (Detection+Segmentation)	Xilinx	✓		

# Out-of-box Supported Boards

- > ZCU102
- > ZCU104
- > Avnet Ultra96
- > Z7020 SOM
- > ZU2 PCIe board
- > ZU2 SOM
- > ZU9 PCIe Card



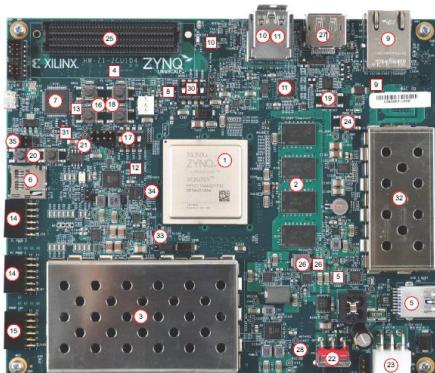
Ultra96



Z7020 SOM



ZU2 SOM



ZCU104

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ZU2 Card



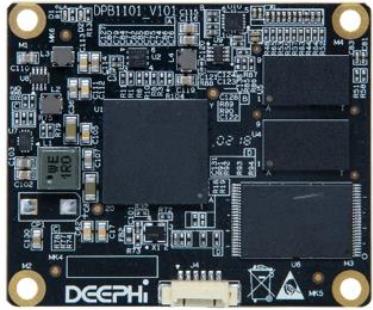
ZU9 Card



ZCU102ektronik  
Your Reliable Supplier

XILINX

# Video Surveillance ML Solutions



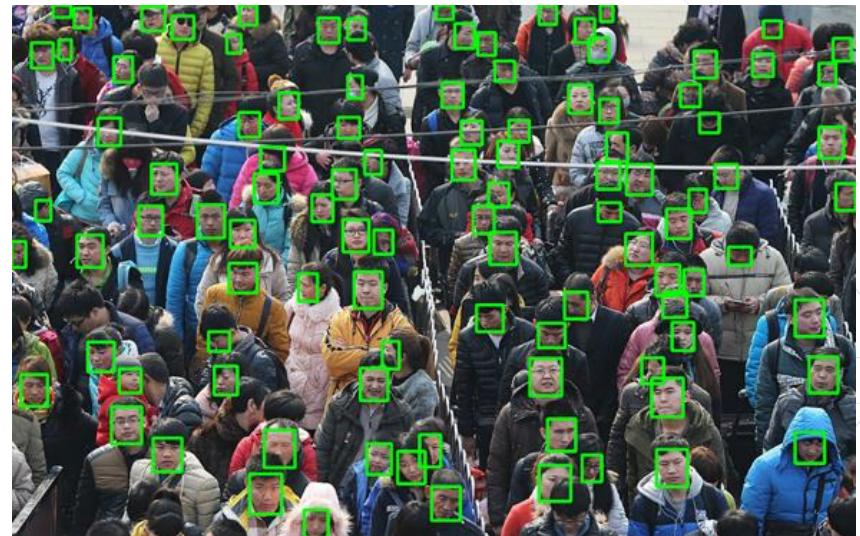
Intelligent  
IP Camera Solution

Face recognition camera  
with Zynq7020

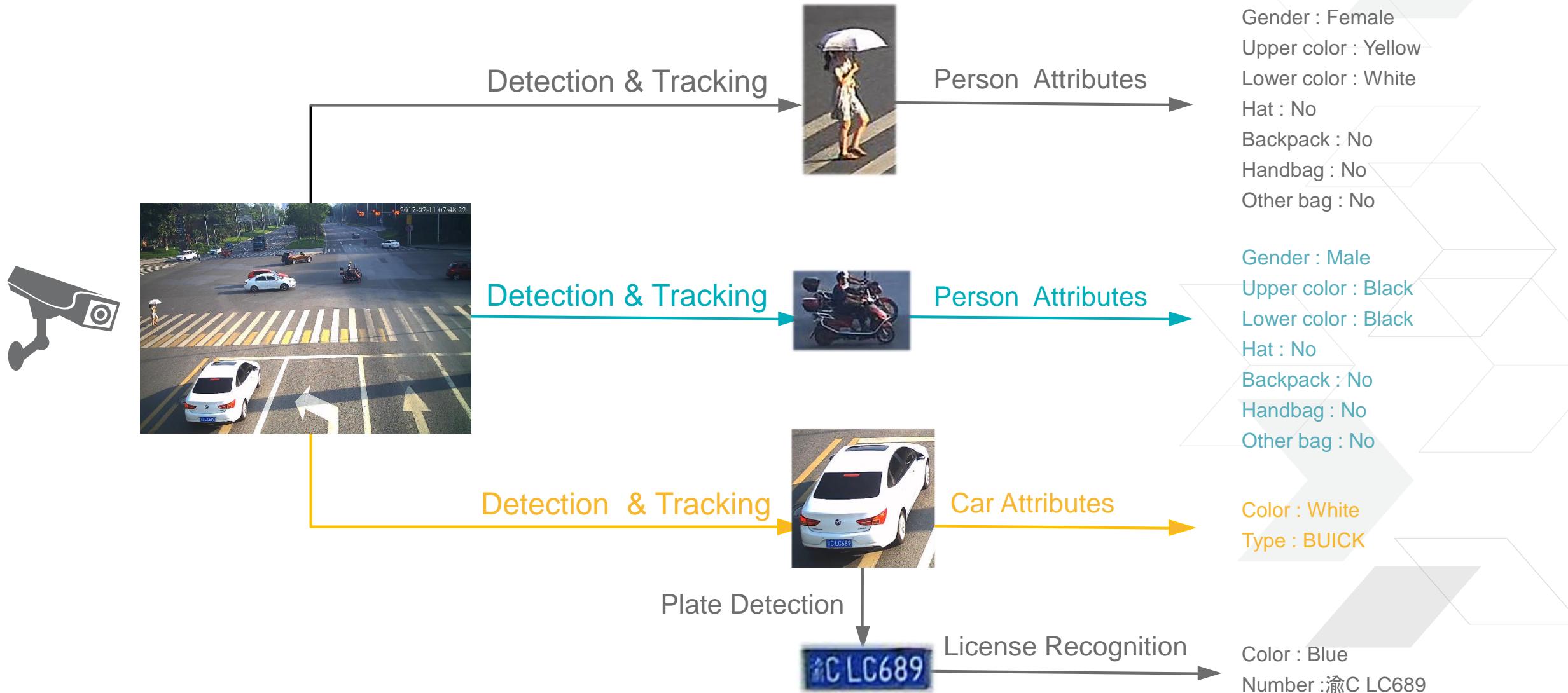


Video Analytics  
Acceleration Solution

12-channel 1080P Video Analytics  
with ZU9EG



# Video Surveillance ML Ref Design

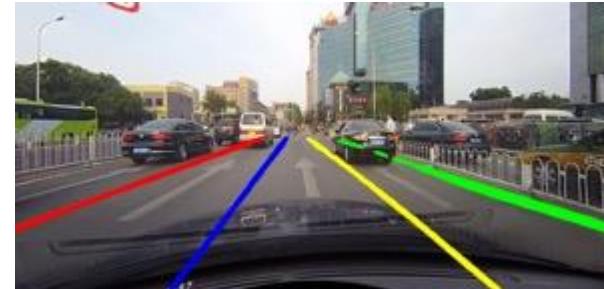


# ADAS/AD ML Reference Design

2D/3D Object Detection



Lane Detection



Pedestrian Detection



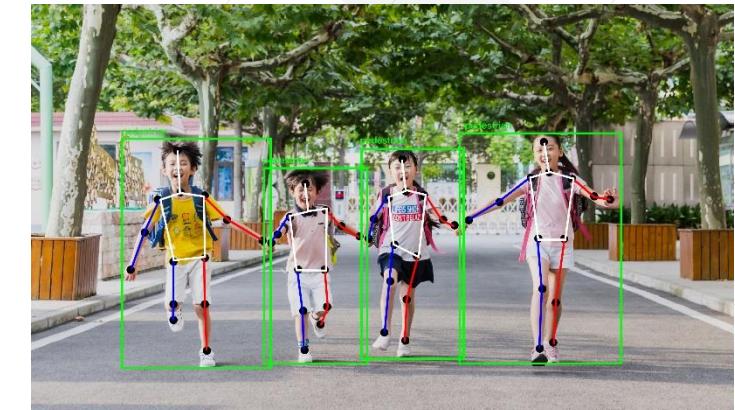
Segmentation + Detection



Segmentation



Pose Estimation



# 8CH Detection Demo

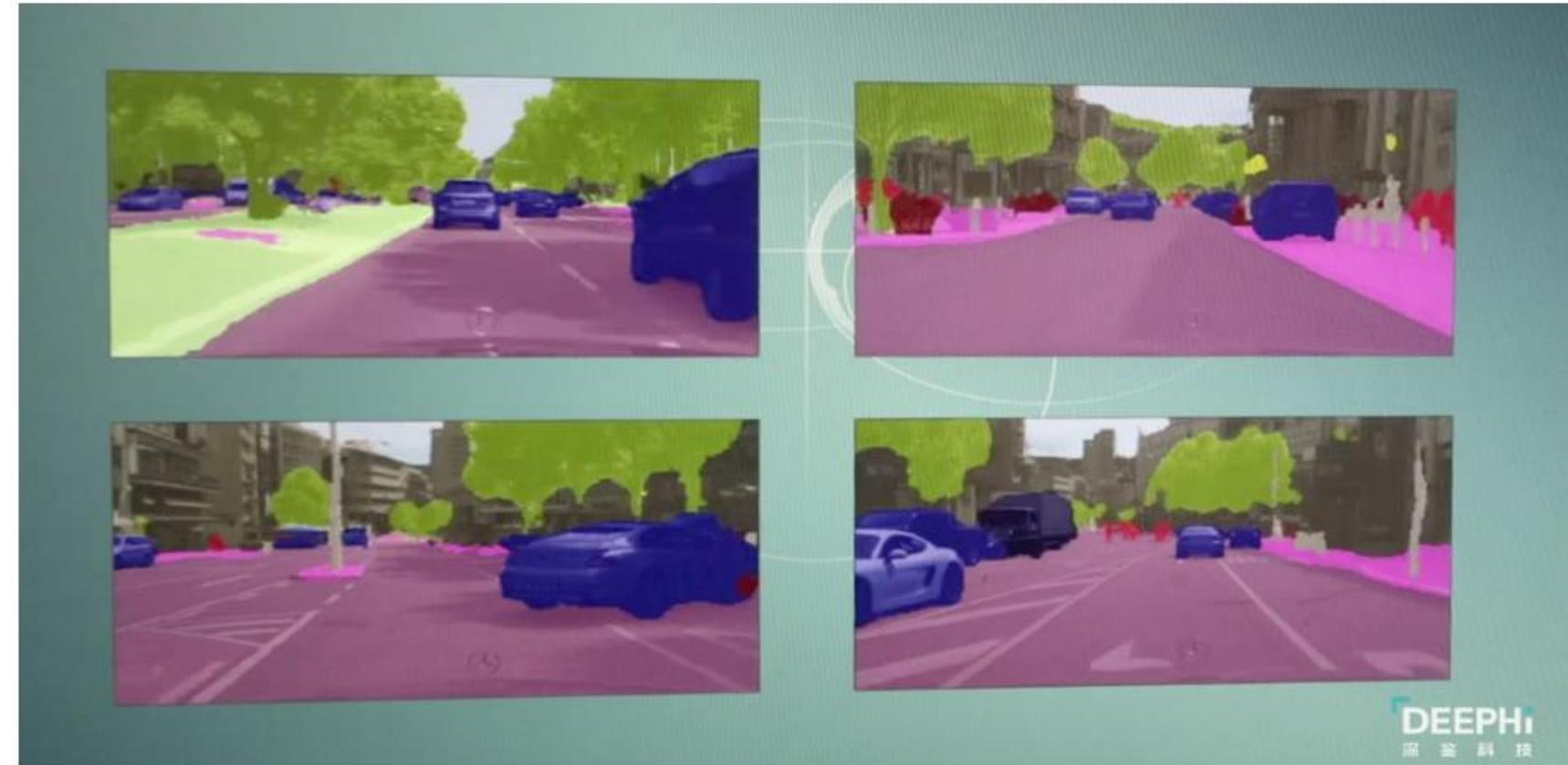
- > Xilinx device
  - >> ZU9EG
- > Network
  - >> SSD compact version
- > Input image size to DPU
  - >> 480 \* 360
- > Operations per frame
  - >> 4.9G
- > Performance
  - >> 30fps per channel



\*Removed Video

# 4-ch Segmentation + Detection Demo

- > Xilinx device
  - >> ZU9EG
- > Network
  - >> FPN compact version
  - >> SSD compact version
- > Input image size to DPU
  - >> FPN – 512 \* 256
  - >> SSD – 480 \* 360
- > Operations per frame
  - >> FPN – 9G
  - >> SSD – 4.9G
- > Performance
  - >> 15fps per channel

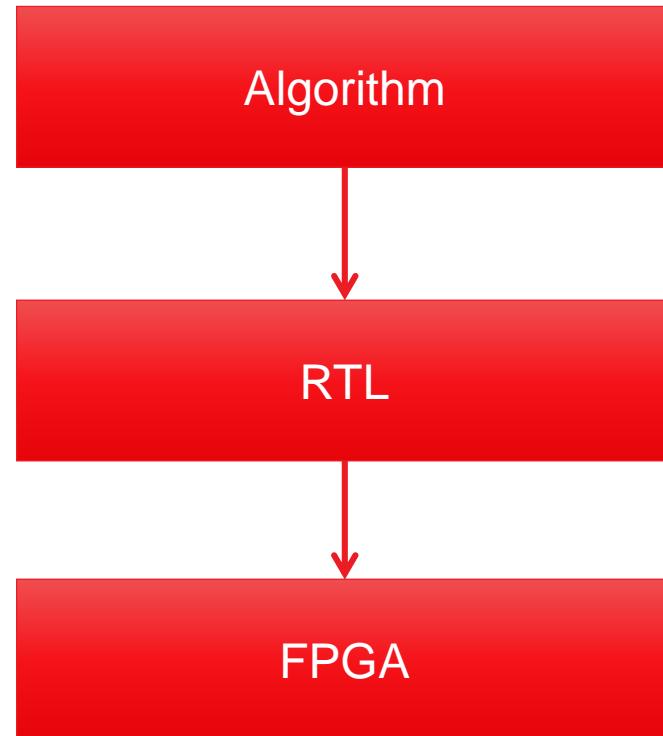


\*Removed Video

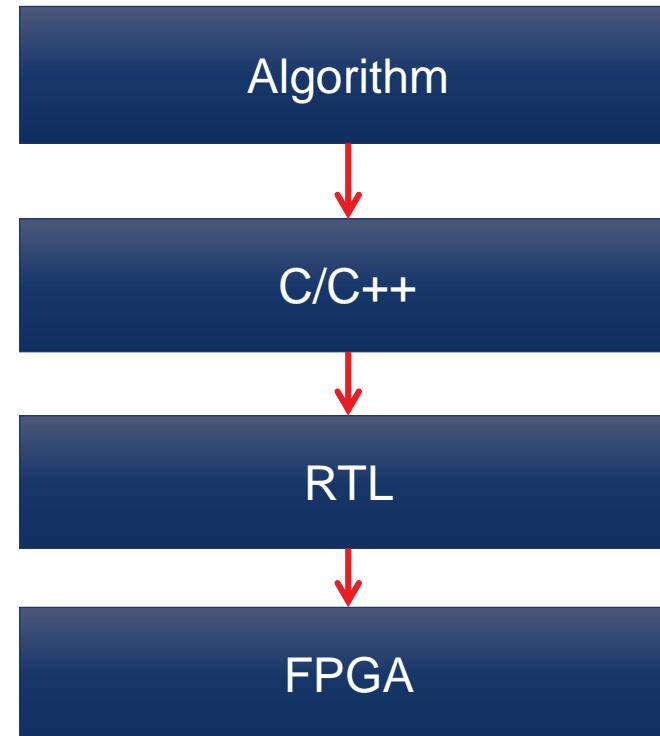
# ML Development with DPU/DNNDK



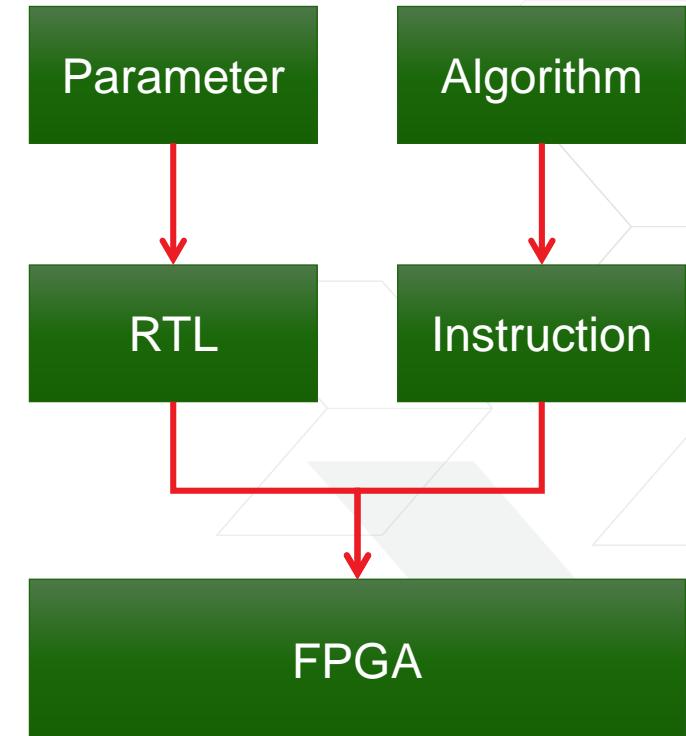
# Development Method



Traditional



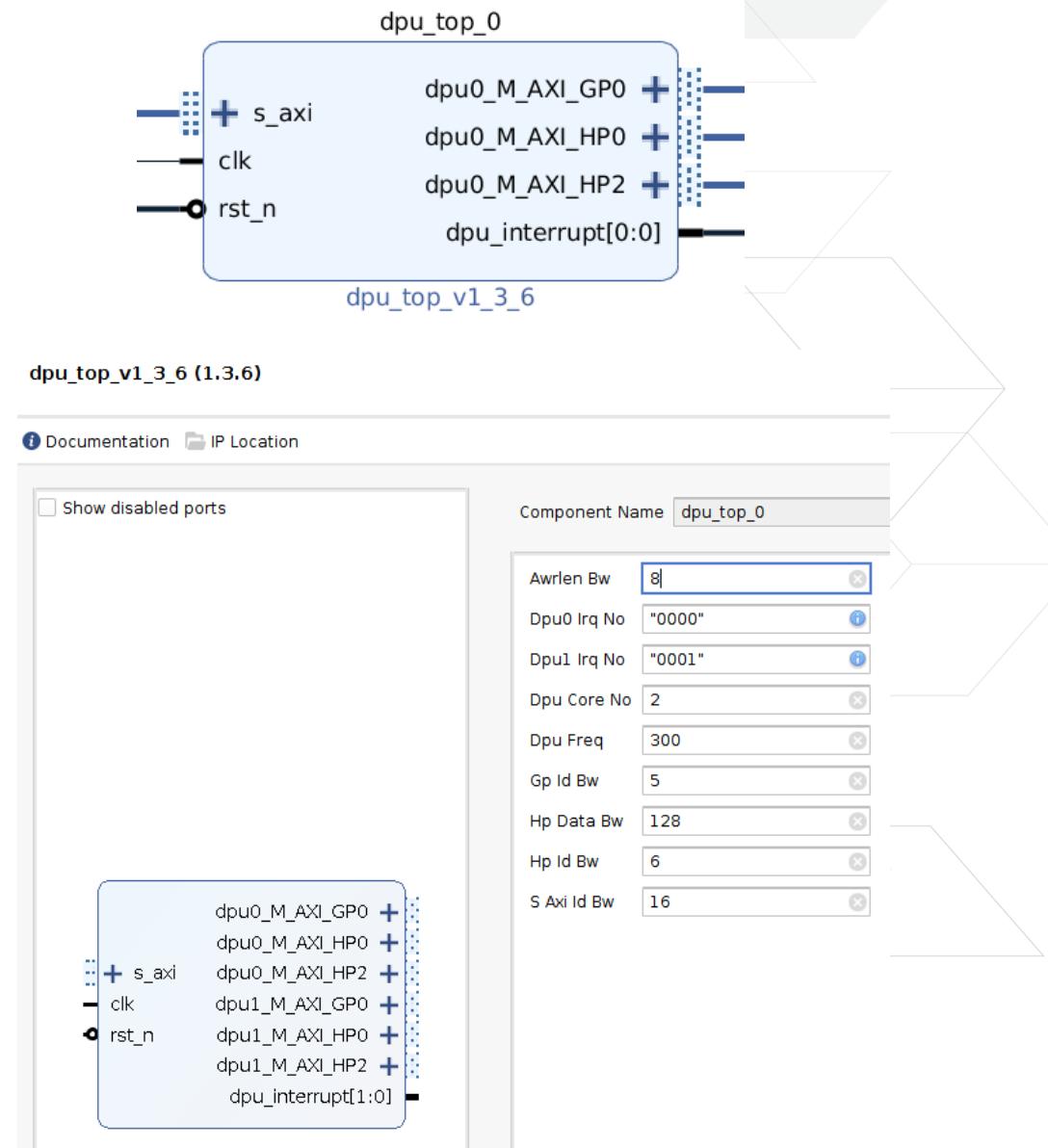
OpenCL/HLS



Xilinx

# HW Integration with Vivado IPI

- > Add DPU IP into repository
- > Add DPU into block design
- > Configure DPU parameters
- > Connect DPU with MPSoC(for reference)
  - >> M\_AXI\_HP0 <-> S\_AXI\_HP0\_FPD (ZYNQ)
  - >> M\_AXI\_HP2 <-> S\_AXI\_HP1\_FPD (ZYNQ)
  - >> M\_Axi\_GP0 <-> S\_AXI\_LPD(ZYNQ)
  - >> s\_axi <-> M\_AXI\_HPM0\_LPD (ZYNQ)
- > Assign Reg address for DPU in address editor
  - >> e.g. 0x80000000, 4K space for one DPU
- > Create top wrapper
- > Generate bitstream
- > Generate BOOT.BIN using Petalinux etc.



# SW Integration with SDK

## > Device tree configuration

- >> set interrupt number according to block design
- >> set core-num

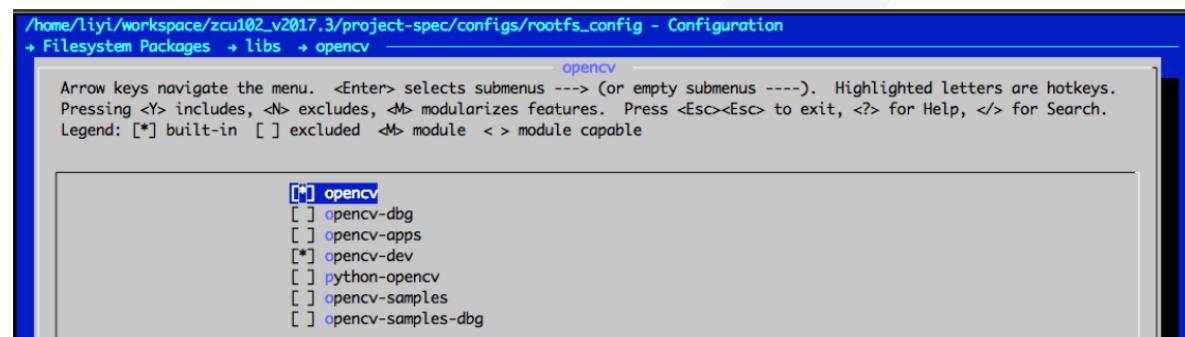
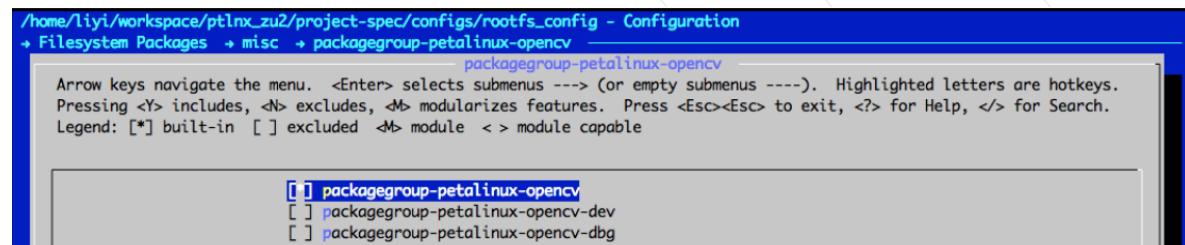
## > OpenCV configuration

- >> Enable in Filesystem Packages -> misc or libs

## > Driver and DNNDK lib

- >> Provide kernel information & OpenCV version to Xilinx
- >> Xilinx will provide driver and DNNDK package with install script
- >> Install driver and DNNDK lib

```
amba {  
...  
dpu@80000000 {  
  
    compatible = "deephisi, dpu";  
    interrupt-parent = <&intc>;  
    interrupts = <0x0 106 0x1 0x0 107 0x1>;  
    reg = <0x0 0x80000000 0x0 0x700>;  
    memory = <0x60000000 0x20000000>;  
    core-num = <0x2>;  
};  
....  
}
```



# Availability



# Basic and Professional Editions

## > Public Access Timeframe

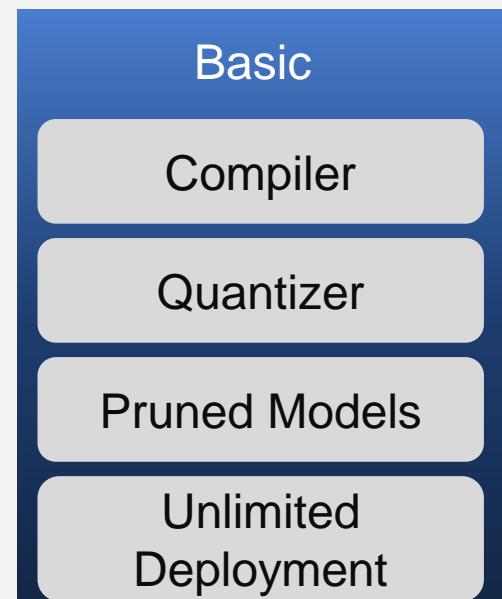
- >> Basic: Now
- >> Basic with Tensorflow: Apr 2019
- >> Professional: May 2019

## > Basic in AWS Cloud – Apr 2019

## > Add-on design service – SoW

Free

Everything you need  
to do it yourself



## Professional

3-day On-site  
Training

Pruning Tools

Compiler

Quantizer

Pruned Models

Unlimited  
Deployment

★ For Professional Edition pricing,  
please inquiry Xilinx AI marketing

Access Pruning Technology  
&  
3-day on-site training by a top-  
notch ML expert  
&  
30-day evaluation with encrypted  
pruning output

# Availability

## > DNNDK & DPU

- >> [DNNDK basic edition - Download from Xilinx.com](#)
- >> Pruning tool, separate upon request
- >> DPU available for evaluation & system integration upon request

## > Demos & Ref Designs

- >> General: Resnet50, Googlenet, VGG16, SSD, Yolo v2/v3, Tiny Yolo v2/v3, Mobilenet v1/v2 etc..
- >> Video surveillance: face detection & traffic structure
- >> ADAS/AD: multi-channel detection & segmentation
- >> DPU TRD (Work in progress)

## > Documentation

- >> [DNNDK user guide – UG1327](#)
- >> [DNNDK for SDSoC user guide – UG1331](#)
- >> Edge AI tutorials - <https://github.com/Xilinx/Edge-AI-Platform-Tutorials>
- >> DPU product guide & tutorial (Work in progress)

## > Request or Inquiry

- >> Please contact Andy Luo, [andy.luo@xilinx.com](mailto:andy.luo@xilinx.com)

**Adaptable.  
Intelligent.**

