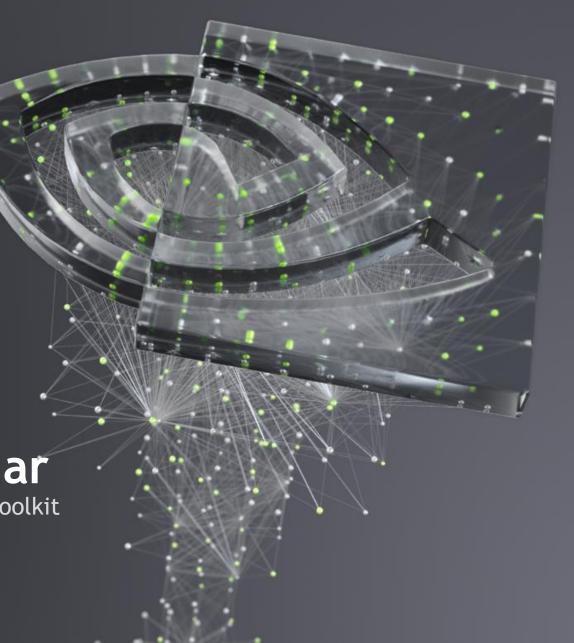


NVIDIA Hands-on Seminar

Deepstream SDK, TensorRT, Transfer Learning Toolkit

Jonghwan Lee, Deep Learning Data Scientist, NVIDIA

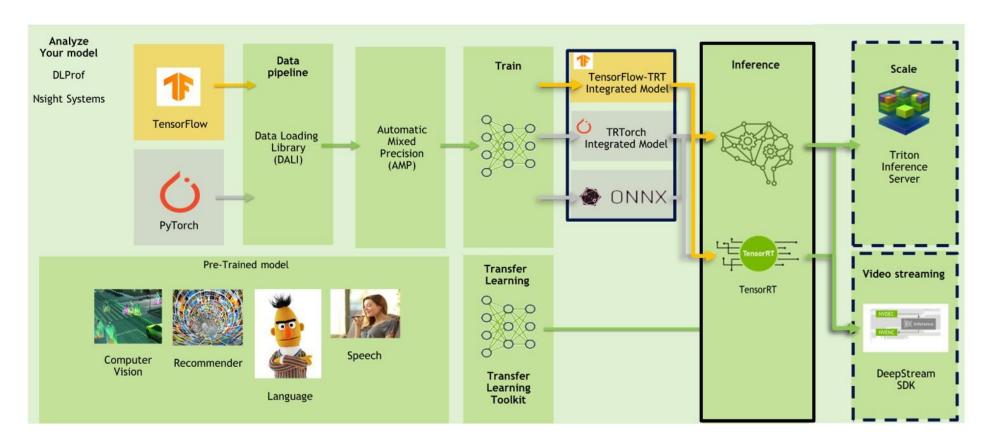






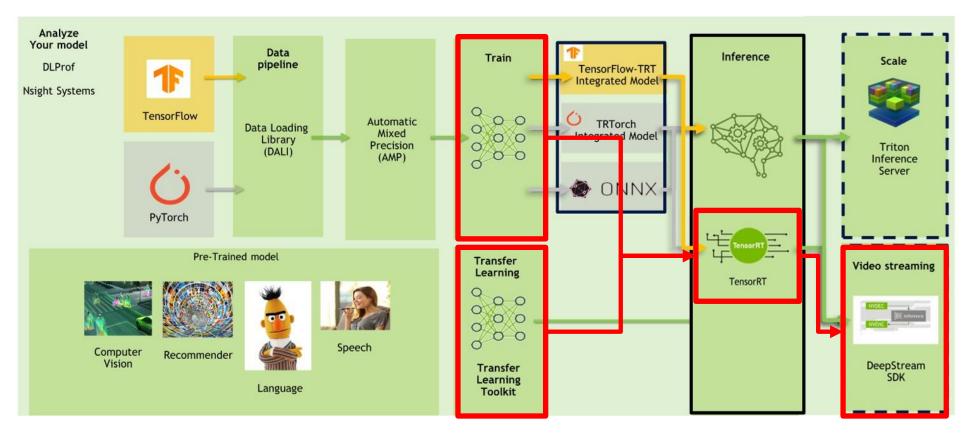
NVIDIA INFERENCE SOLUTIONS

Easily deploy AI, Tools and Libraries to accelerate your AI model and Service



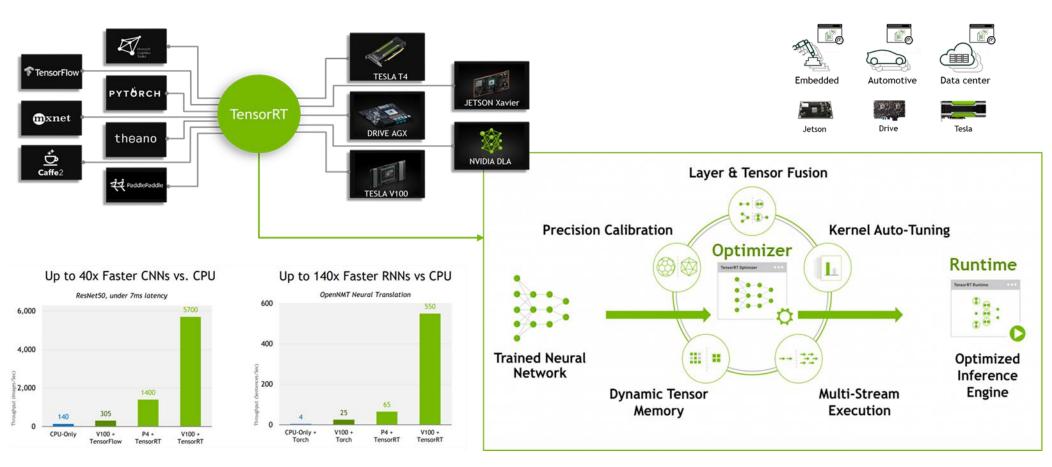
NVIDIA INFERENCE SOLUTIONS

Easily deploy AI, Tools and Libraries to accelerate your AI model and Service



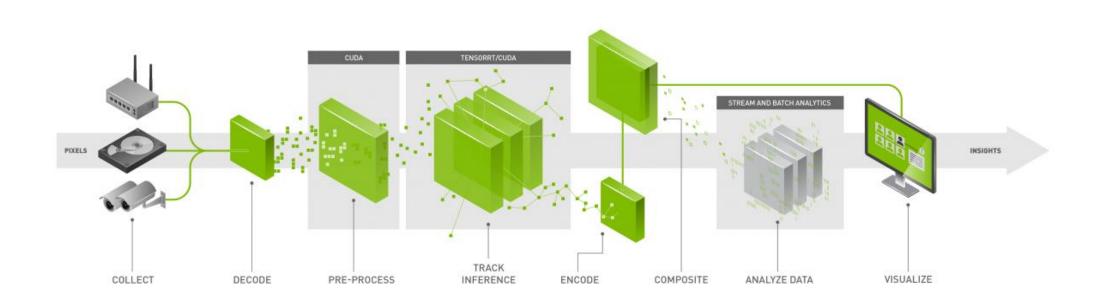
NVIDIA TensorRT

From Every Framework, Optimized For each Target platform



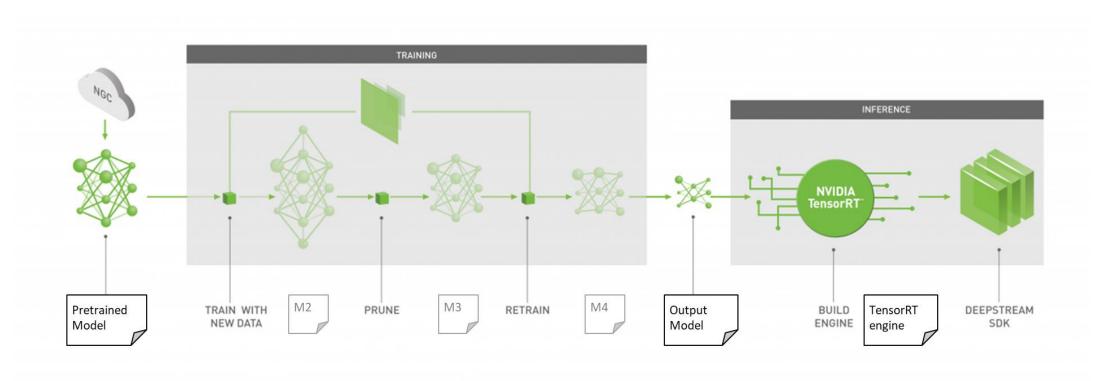
NVIDIA Deepstream SDK

Analyze Data from Cameras, Sensors and IoT Gateways in Real-Time



NVIDIA Transfer Learning Toolkit

Accelerates Throughput on Leading Industry Platforms



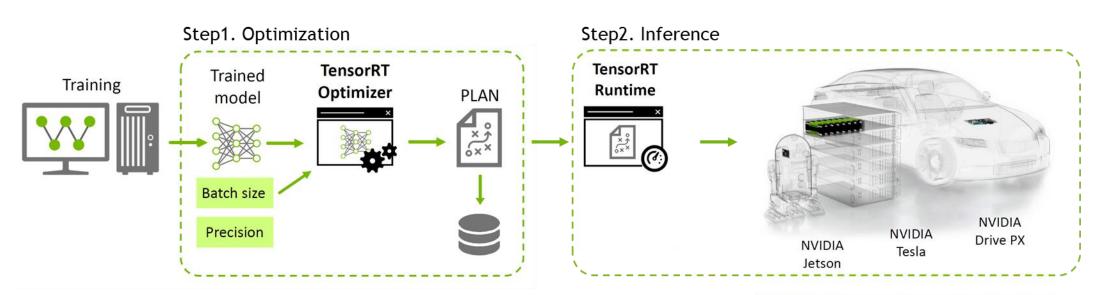
1. Download docker container -> 2. Pull Model -> 3. Train with your data -> 4. Prune -> 5. Retrain -> 6. Export



TENSORRT WORKFLOW



How to accelerate DNN inference with TensorRT



Import model → Optimize model → Serialize plan

Deserialize plan → Build Runtime engine → Deploy Runtime

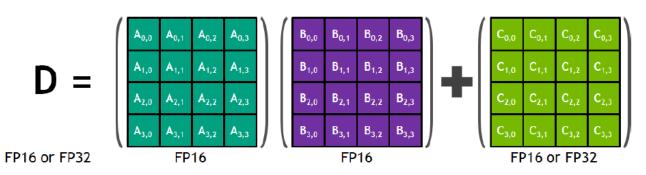
NVIDIA TENSOR CORE

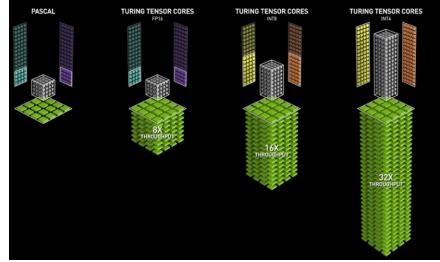


Mixed precision Matrix-Multiply-and-Add

Ampere has new 3rd TensorCore

- Available in Volta and Turing architecture GPUs
- ► 125 Tflops in FP16 vs 15.7 Tflops in FP32 (8x speed-up)
- Optimized 4x4x4 dot operation (GEMM)

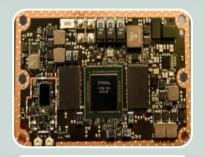


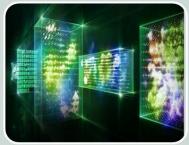


NVDLA



NVIDIA Deep Learning Accelerator









Developed as part of Xavier

- NVIDIA's

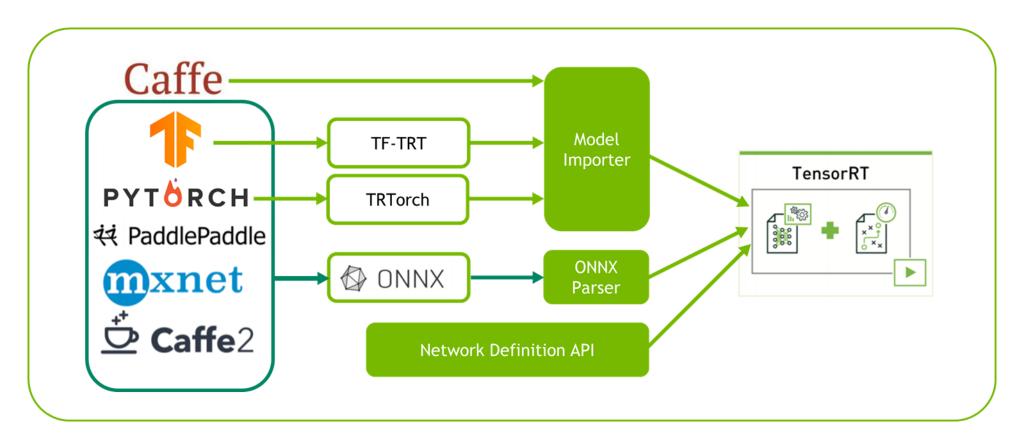
SOC for autonomous driving applications

Optimized for Convolutional Neural Networks (CNNs), computer vision Open source architecture and RTL release. NVDLA Compiler is now open source in the latest NVDLA v1.2.0-OC release Complete solution with Verilog and Cmodel, Linux drivers, test benches, kernel and user-mode software and virtual platform

NVIDIA TENSORRT



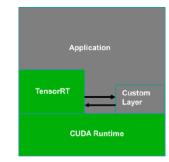
TensorRT Model import flow

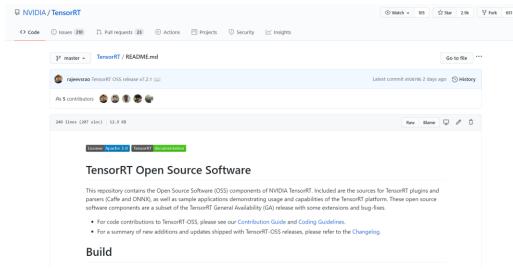


NVIDIA TENSORRT OSS

Supports custom plug-ins by CUDA

- Allow users to express and provide implementations of novel layers
 - · TensorRT provides APIs and dedicated implementations for most common layers
 - Use the custom layer API for infrequent or more innovative layers or your own confident implementation
 - · Register custom implementations via a callback mechanism
 - · Can be used in conjunction with reduced precision optimization

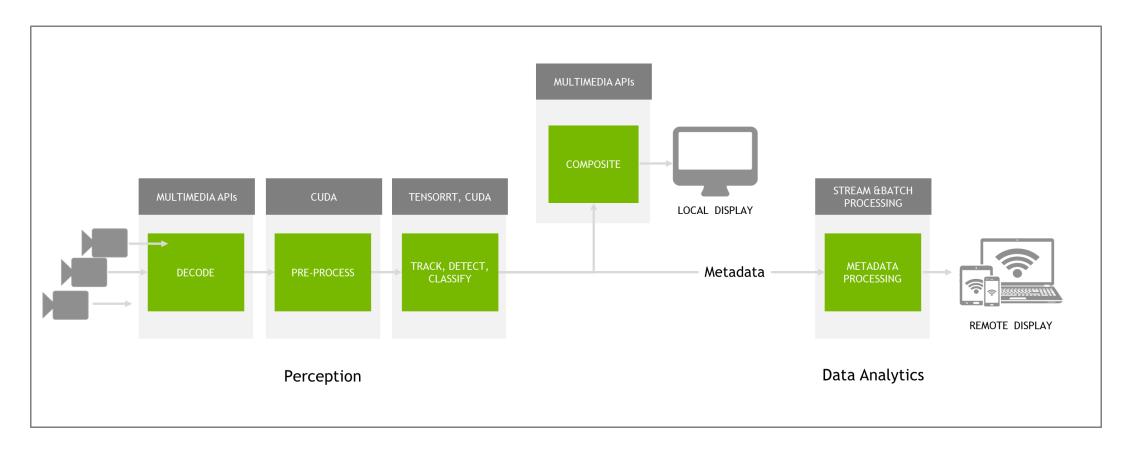




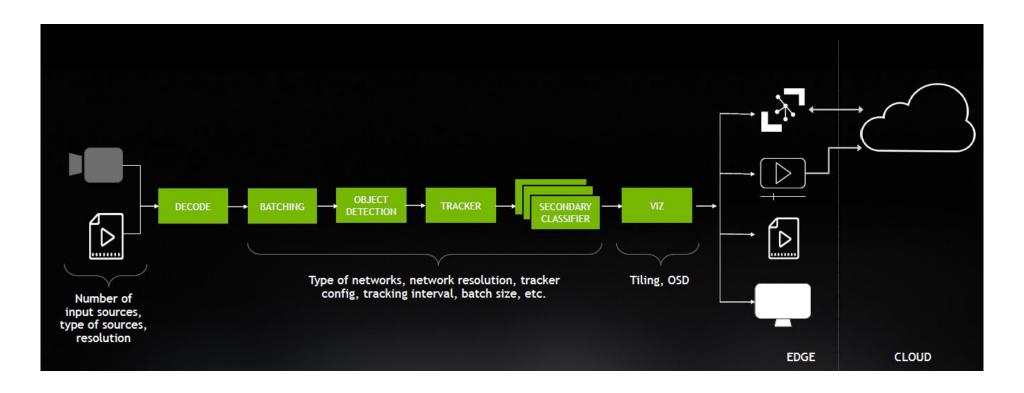


GENERAL INTELLIGENCE VIDEO ANALYSIS

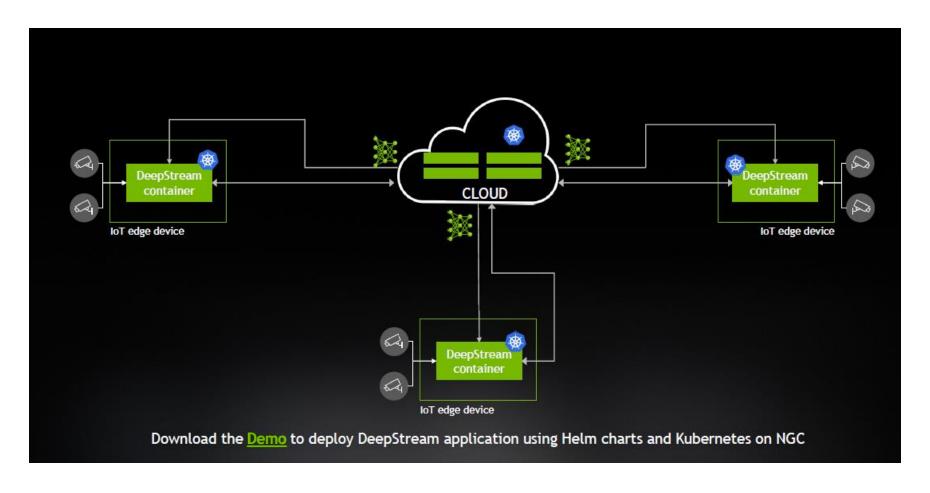
Framework for analyzing video



End-to-End Deepstream Application - deepstream-test5

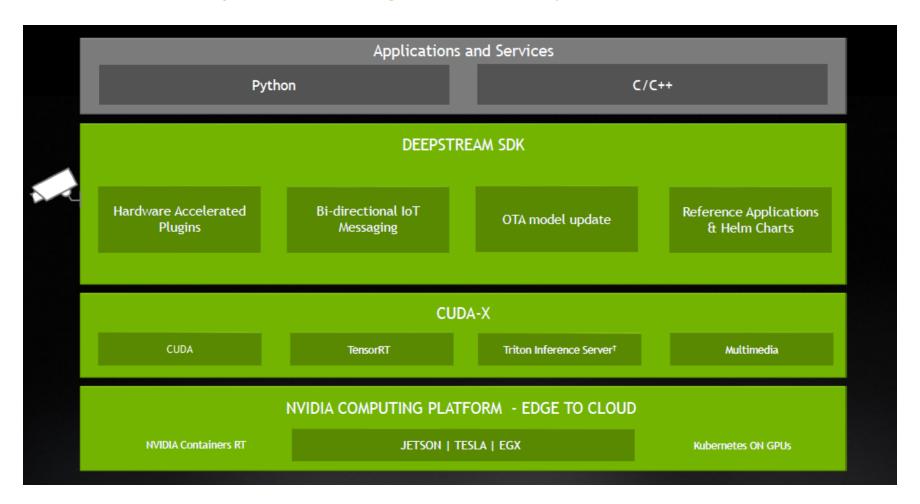


At scale deployment



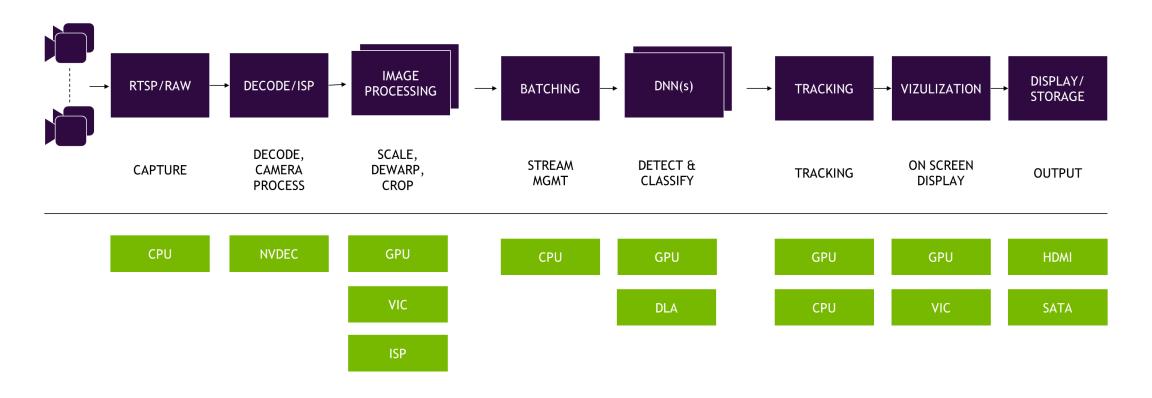
WHAT IS DEEPSTREAM?

AI-powered Intelligent Video Analytics framework



DEEPSTREAM GRAPH ARCHITECTURE

AI-powered Intelligent Video Analytics framework



Gstreamer Foundations

The DeepStream SDK is based on the open source <u>GStreamer multimedia framework</u>. There are a few key concepts in GStreamer that we need to touch on before getting started. These include Elements, Pads, Buffers, and Caps. We will be describing them at a high level, but encourage those who are interested in the details to read the <u>GStreamer</u> documentation to learn more.



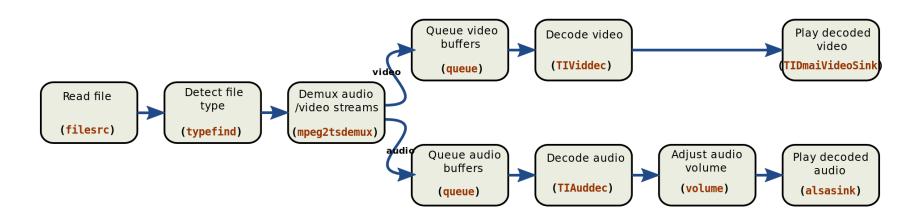
Example of Gstreamer pipeline

The goals of GStreamer are to seprate the application (e.g. Video player, Video editor, etc.) from the streaming media complexity (e.g. hardware acceleration, remoteness) GStreamer - streaming media

D-Bus – inter process communication

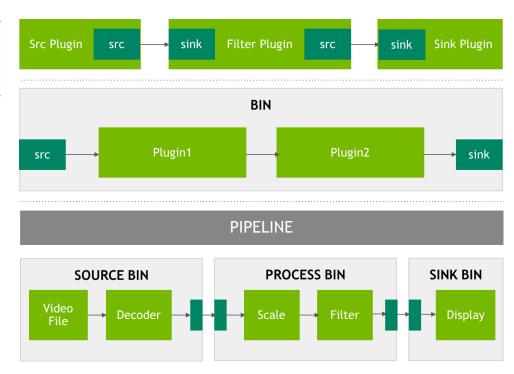
Use gst-launch command to create the GStreamer pipeline

gst-rtsp-server - v4l2src ! video/x-raw,width=1280,height=720 ! omxh264enc ! video/x-h264,profile=baseline ! h264parse config-interval=1 ! rtph264pay name



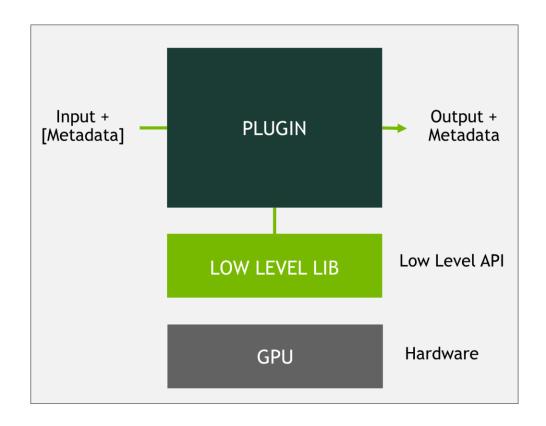
Gstreamer pipeline

Level	Component	Function
1	PLUGINS	Basic building block connected through PADs
2	BINS	A container for a collection of plugins
3	PIPELINE	Top level bin providing a bus and managing the synchronization



Deepstream Building block

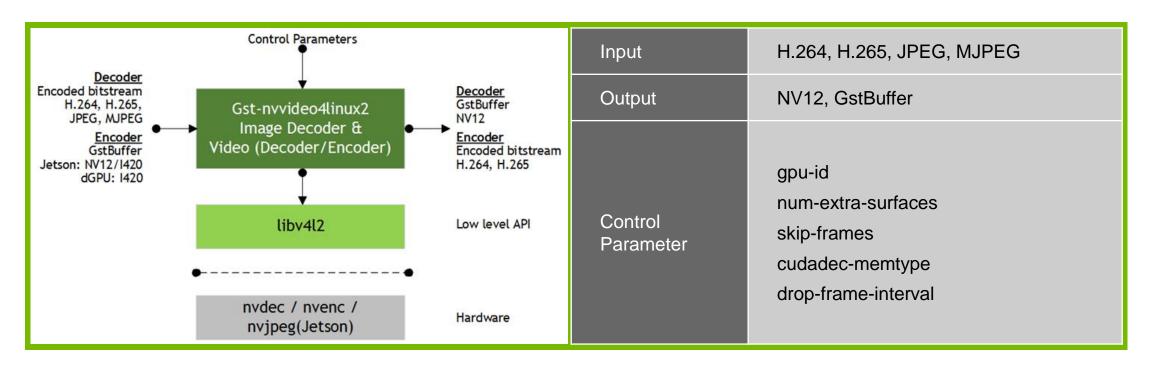
- A plugin model-based pipeline architecture
- Graph-based pipeline interface to allow high-level component interconnect
- Heterogenous processing on GPU and CPU
- Hides parallelization and synchronization under the hood
- Inherently multi-threaded



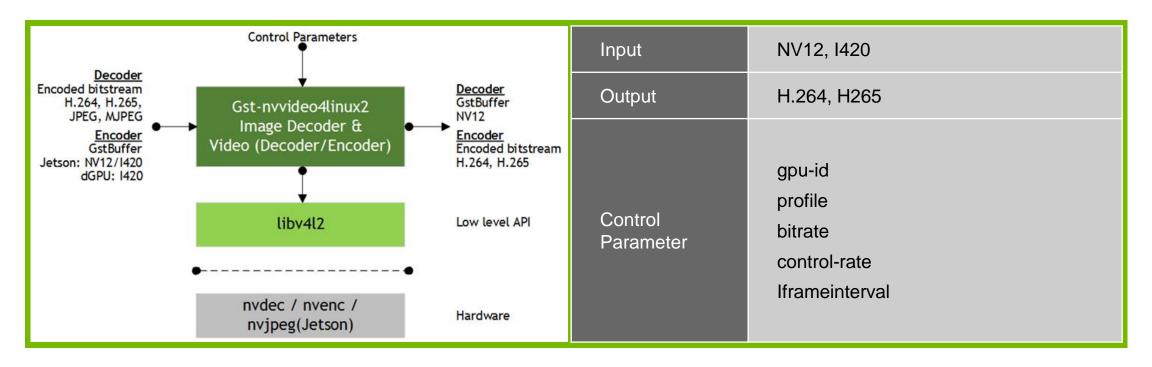
Deepstream accelerated plugin

Plugin Name	Functionality
Gst-nvinfer	DL inference for detection, classification and segmentation with TRT engine
Gst-nvinferserver	DL inference for detection, classification and segmentation with any other native framework
Gst-nvtracker	Reference object trackers; KLT, IOU, NvDCF
Gst-nvstreammux	Stream aggregation, multiplexing and batching
Gst-nvstreamdemux	Demux batched frames into individual buffers
Gst-nvmultistreamtiler	Composites a 2D tile from batched buffers
Gst-nvdsosd	Draw boxes and text overlay
Gst-nvvideoconvert	Scaling, format conversion and rotation
Gst-nvdewarper	Dewarping for fish-eye degree cameras
Gst-nvof	Fast optical flow calculation on special HW accelerator for it
Gst-nvofvisual	Visualize input optical flow vector on frames
Gst-nvsegvisual	Visualize segmentation results on frames
Gst-nvvideo4linux2	Hardware accelerated decode and encode(NVDEC / NVENC)
Gst-nvjpegdec	Decoding JPEG images
Gst-nvmsgconv	Metadata generation
Gst-nvmsgbroker	Messaging to cloud
Gst-nvdsanalytics	ROI filtering, Overcrowding Detection, Direction Detection and Line crossing

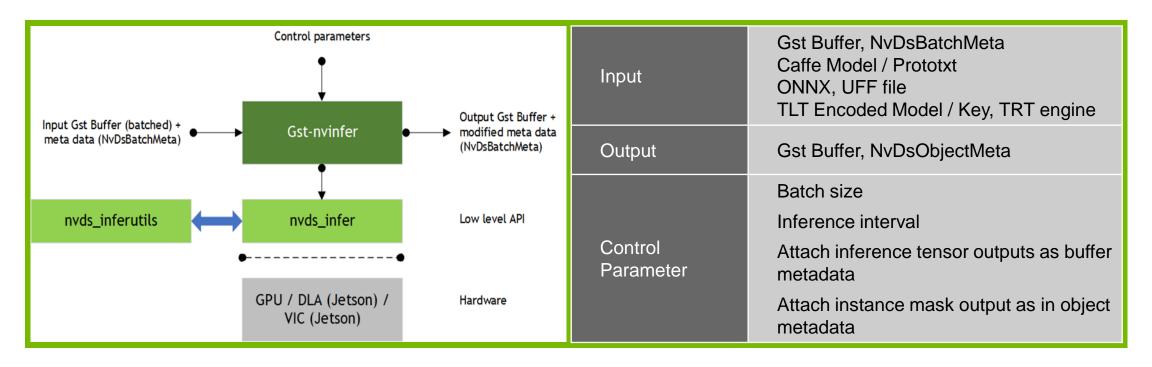
Deepstream accelerated plugin - Gst-nvvideo4linux2 (Decode)



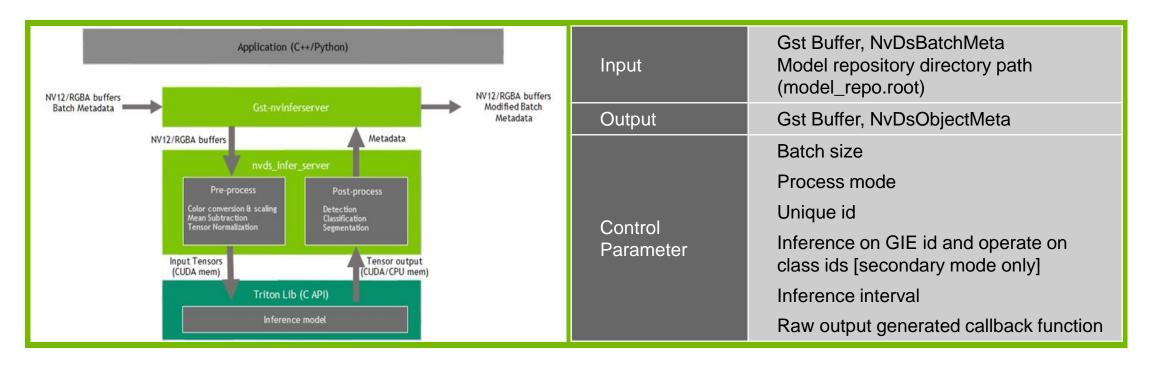
Deepstream accelerated plugin - Gst-nvvideo4linux2 (Encode)



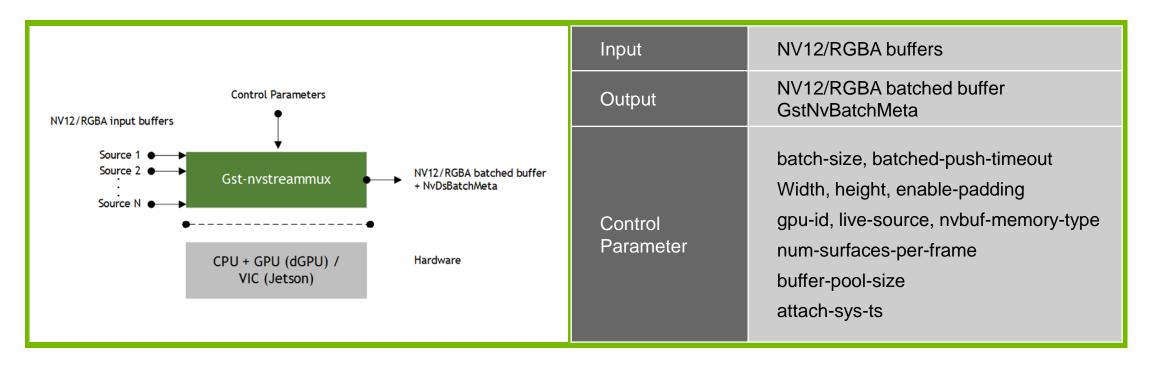
Deepstream accelerated plugin - Gst-nvinfer



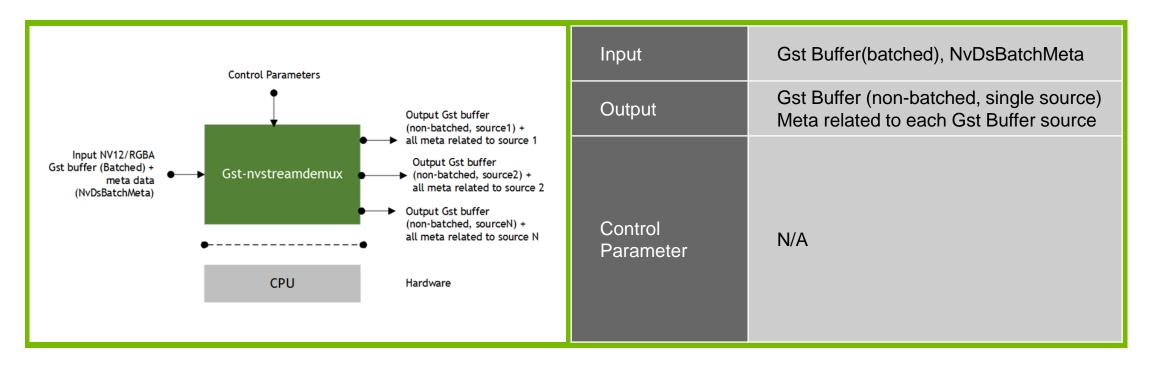
Deepstream accelerated plugin - Gst-nvinferserver



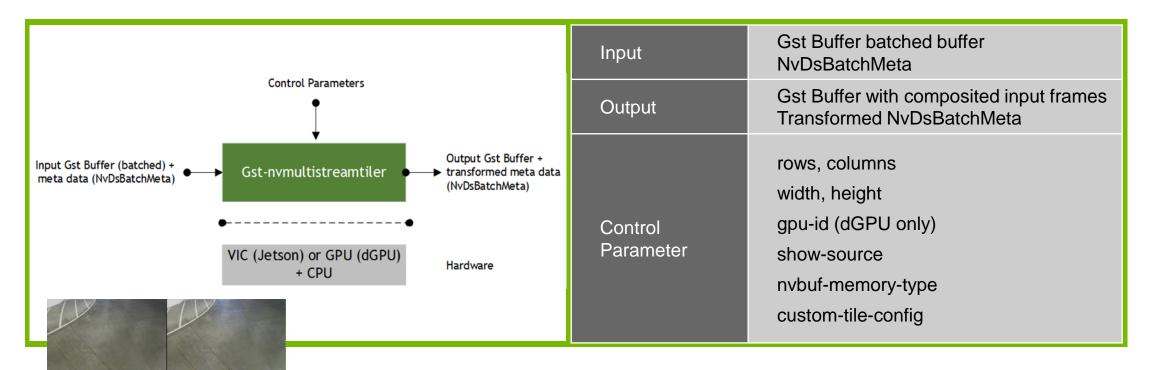
Deepstream accelerated plugin - Gst-nvstreammux



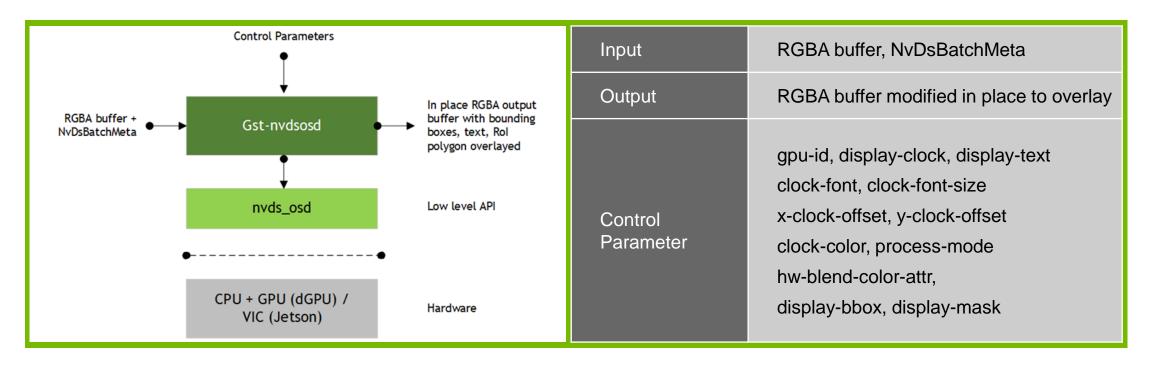
Deepstream accelerated plugin - Gst-nvstreamdemux



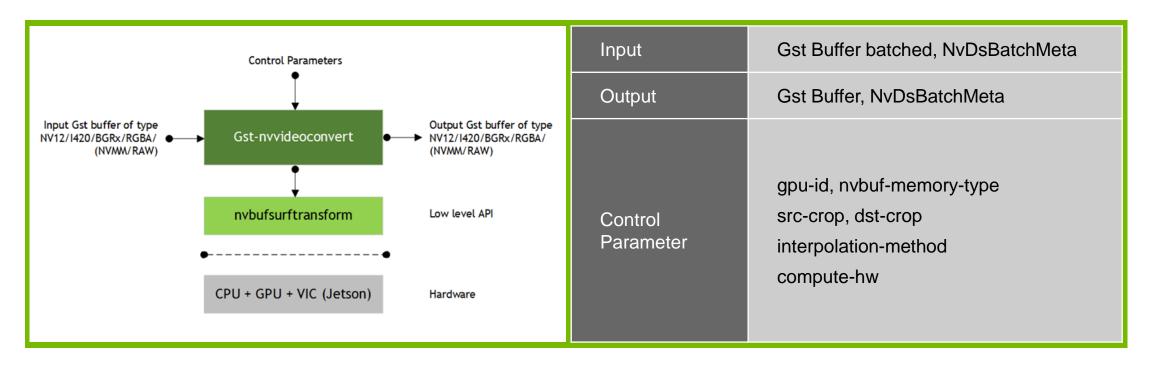
Deepstream accelerated plugin - Gst-nvmultistreamtiler



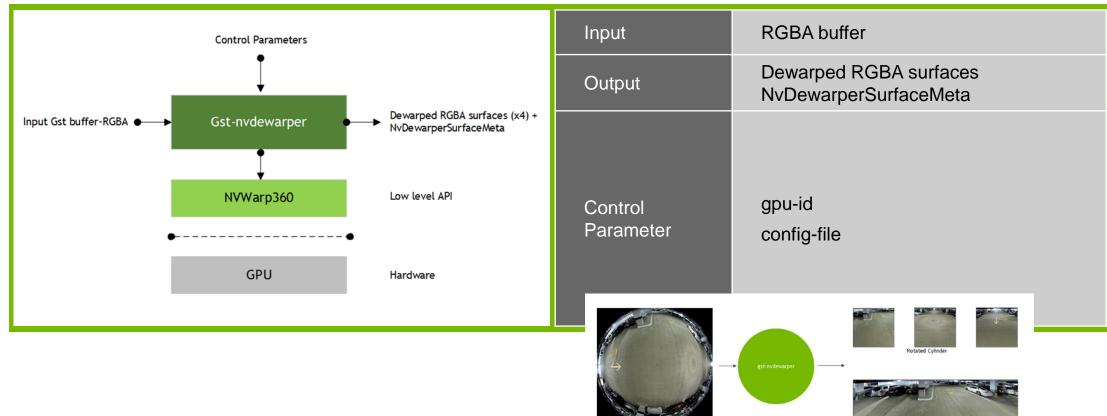
Deepstream accelerated plugin - Gst-nvdsosd



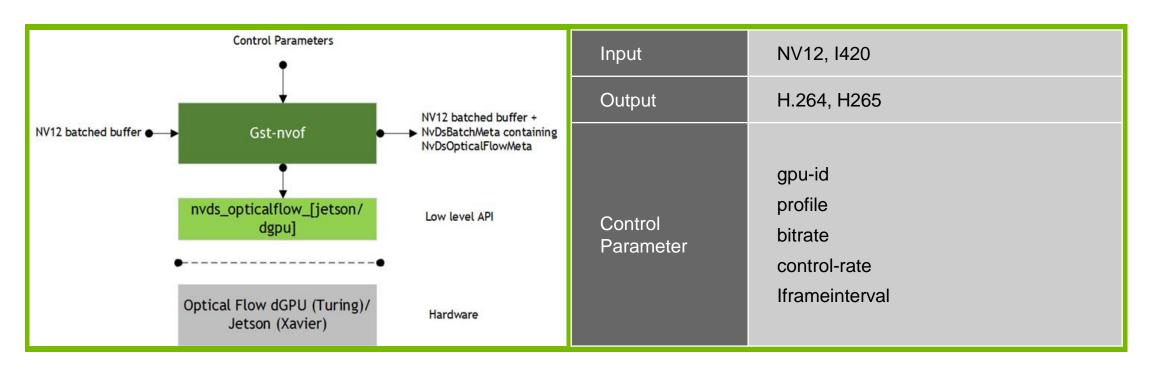
Deepstream accelerated plugin - Gst-nvvideoconvert



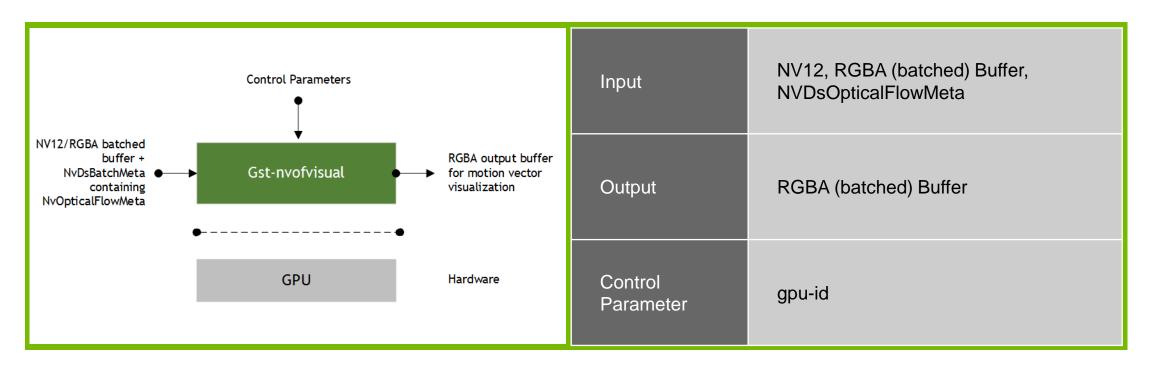
Deepstream accelerated plugin - Gst-nvdewarper



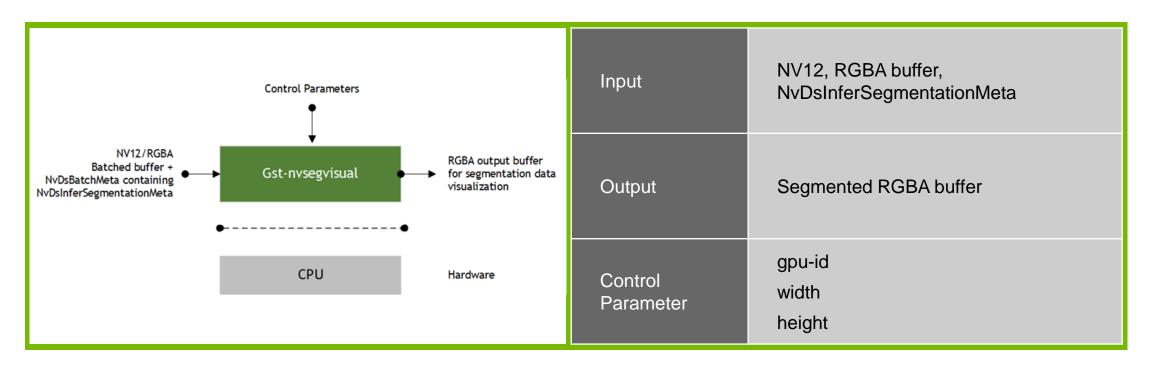
Deepstream accelerated plugin - Gst-nvof



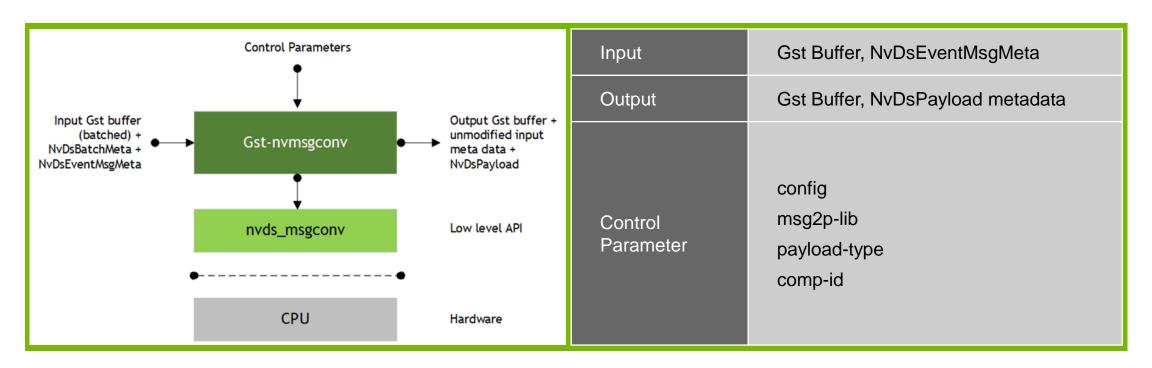
Deepstream accelerated plugin - Gst-nvofvisual



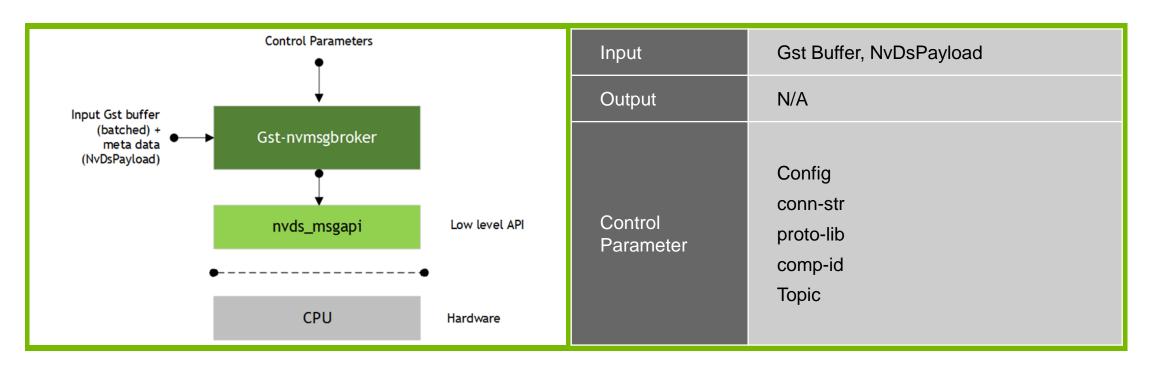
Deepstream accelerated plugin - Gst-nvsegvisual



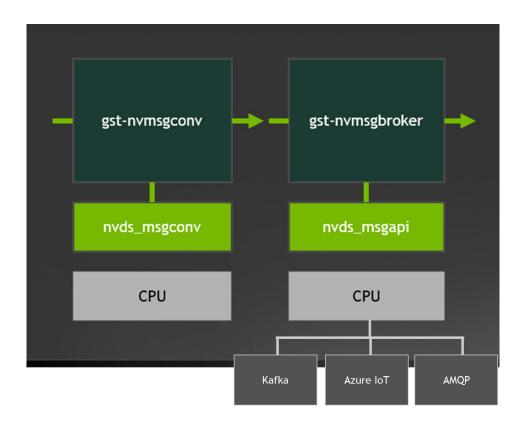
Deepstream accelerated plugin - Gst-nvmsgconv



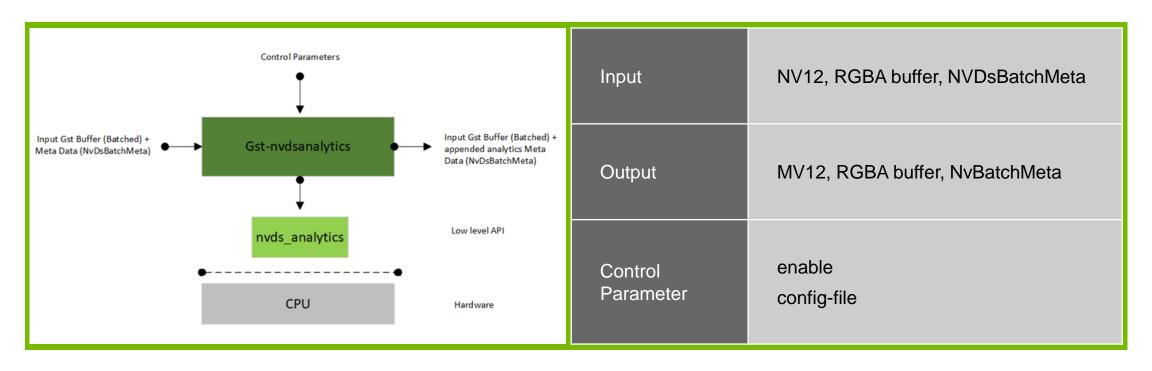
Deepstream accelerated plugin - Gst-nvmsgbroker



Deepstream accelerated plugin - Gst-nvmsgconv & Gst-nvmsgbroker



Deepstream accelerated plugin - Gst-nvdsanalytics



Get started applications (available both C and Python)

Name	Function	
deepstream-test1	DeepStream Hello world. Single video from file to on screen display with bounding box	Decode Batching Object detection
deepstream-test2	Builds on test1 and adds secondary object classification on detected objects	Object detection Classification
deepstream-test3	Builds on test1 and adds multiple video inputs	Decode → Batching → Object detection →
deepstream-test4	Builds on test1 and adds connections to IoT services thru the nvmsgbroker plugin	Object detection Converter Message Broker

Native-C apps: sources/apps/sample apps/

Python apps: $https://github.com/\overline{N}VIDIA-AI-IOT/deepstream python apps$



DEEPSTREAM SDK HANDS-ON

Setting up for the Hands-on

Prerequisites

Host machine

- SSH terminal
- > VLC player
- > (For TLT exercise) GPU machine is required & docker runtime environment
- \$ docker run --gpus all -it -v \$(pwd):/workspace -w /usr/local/src -p 8888:8888 cycoslee/nv-deepstreamsdk-handson:tlt_host_210127

Jetson NX (based on Jetpack 4.4.1)

- Docker container pull
- \$ sudo docker run -ti --runtime=nvidia --rm --net=host -e DISPLAY=\$DISPLAY -w /opt/nvidia/deepstream/deepstream-5.0 --device /dev/video0 -v /tmp/.X11-unix/:/tmp/.X11-unix -v \$(pwd):/workspace cycoslee/nv-deepstreamsdk-handson:ds_nx_210127

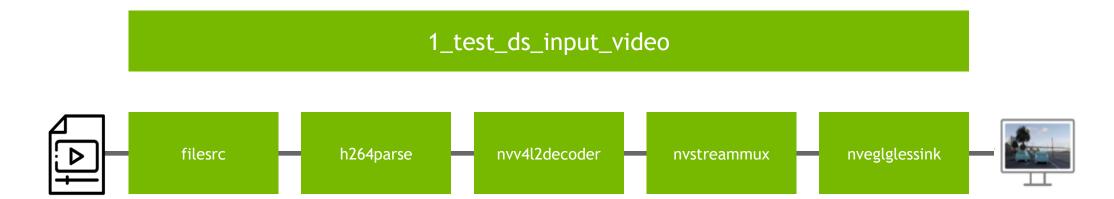
Assets

- IP camera(supports RTSP)
- USB camera(webcam)



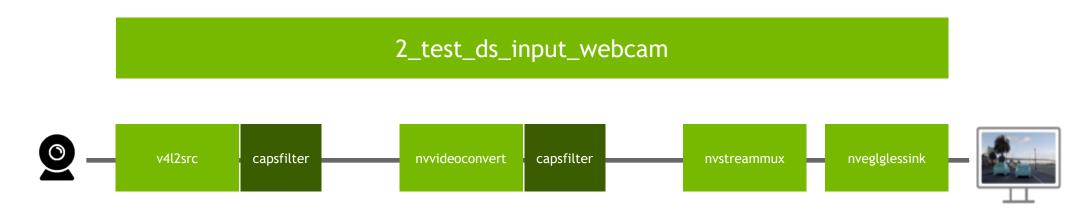
HOW TO CONTROL INPUTS





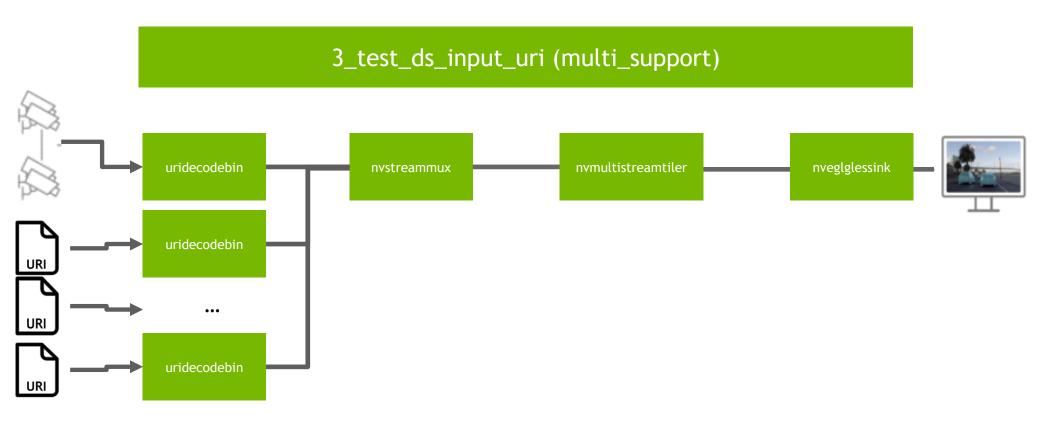
HOW TO CONTROL INPUTS



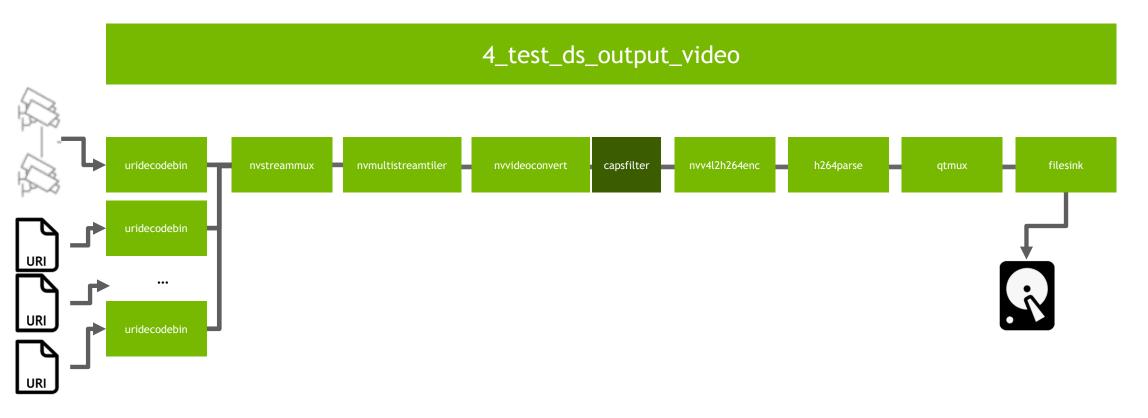




HOW TO CONTROL INPUTS

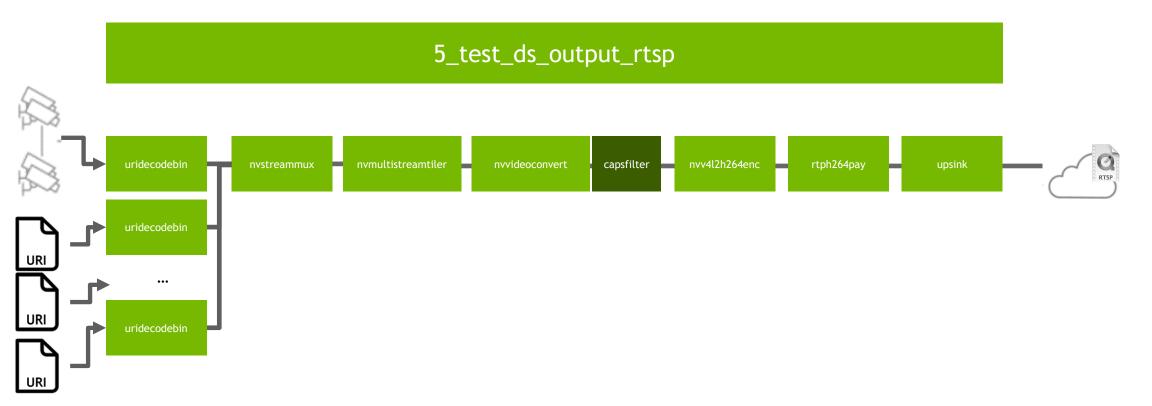


HOW TO CONTROL OUTPUTS





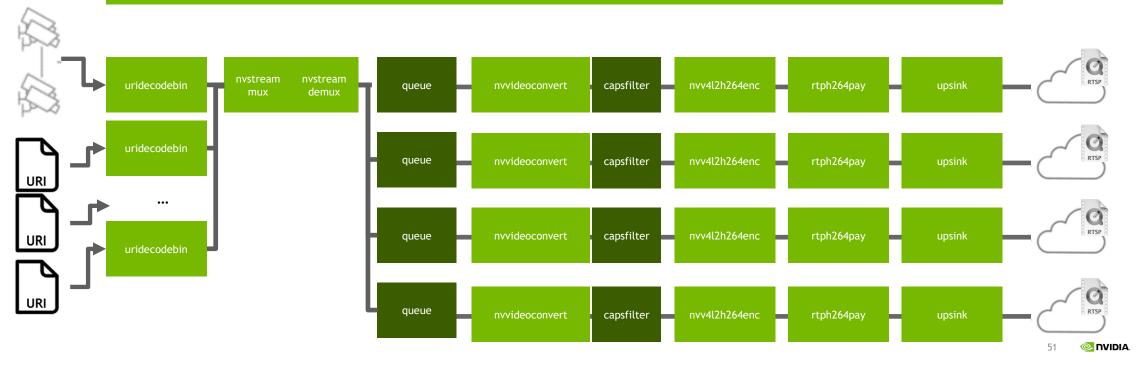
HOW TO CONTROL OUTPUTS



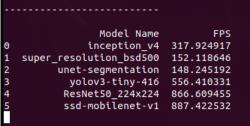
HOW TO CONTROL OUTPUTS



6_test_ds_output_video



How to use TensorRT easily by using trtexec



7_test_trt_benchmark

Benchmarks Targeted for Jetson Xavier NX (Using GPU+2DLA)

The script will run following Benchmarks:

• Names : Input Image Resolution

Inception V4: 299x299ResNet-50: 224x224

OpenPose: 256x456VGG-19: 224x224YOLO-V3: 608x608

• Super Resolution : 481x321

• Unet: 256x256

For benchmark results on all NVIDIA Jetson Products; please have a look at NVIDIA jetson_benchmark webpage

Following scripts are included:

1. Install Requirements for running benchmark script (install_requirements.sh)

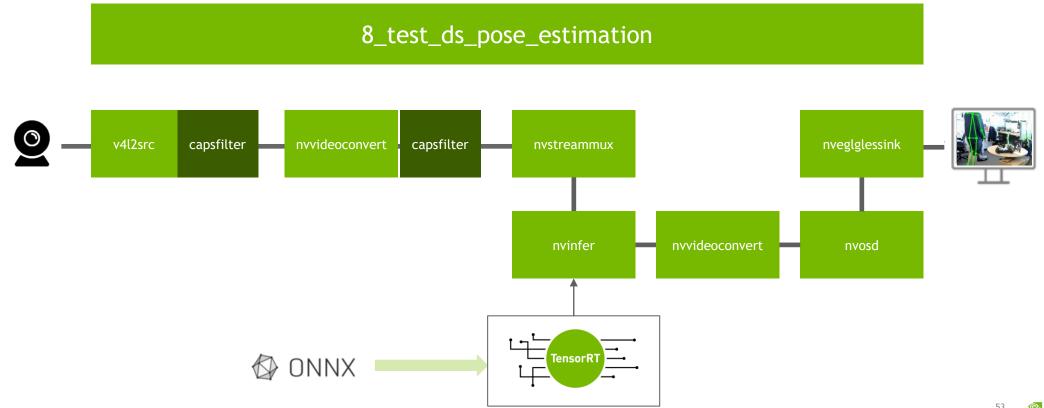
2. CSV files containing parameters (benchmark_csv folder)

3. Download Model (utils/download_models.py)

Model Name	FPS
inception_v4	311.73
vgg19_N2	66.43
super_resolution_bsd500	150.46
unet-segmentation	145.42
pose_estimation	237.1
yolov3-tiny-416	546.69
ResNet50_224x224	824.02
ssd-mobilenet-v1	887.6

By using TensorRT





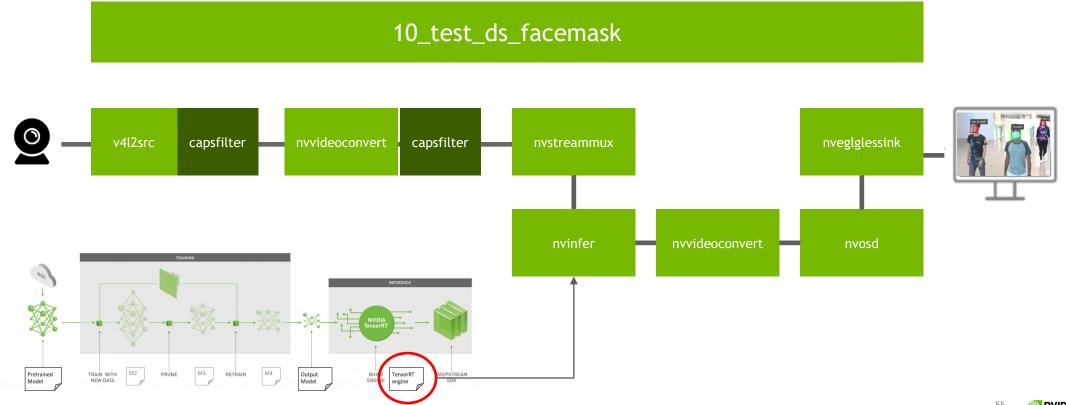
By using TensorRT



9_test_ds_yolov4 SOURCE STREAMMUX **TRACKER DECODE** GIE OSD Darknet ONNX **TensorRT** YOLOv4

By using Transfer Learning Toolkit





Q&A

You can send an E-MAIL me (jonghwanl@nvidia.com) if you have any questions for this

