

AI Box小試身手
加速邊際運算落地運用

奕瑞科技AI研發部門

主講人

工程師 吳宇鴻、謝明原





01. Xavier 介紹

02. Xavier NX 介紹

03. 場景應用

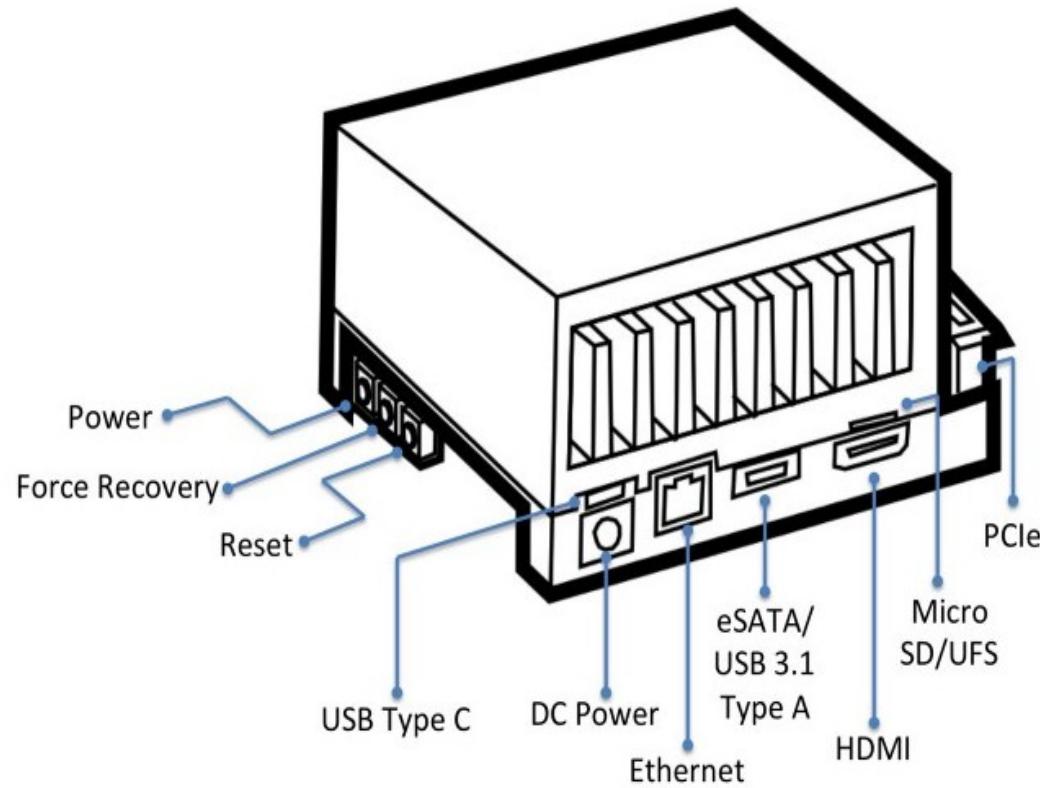
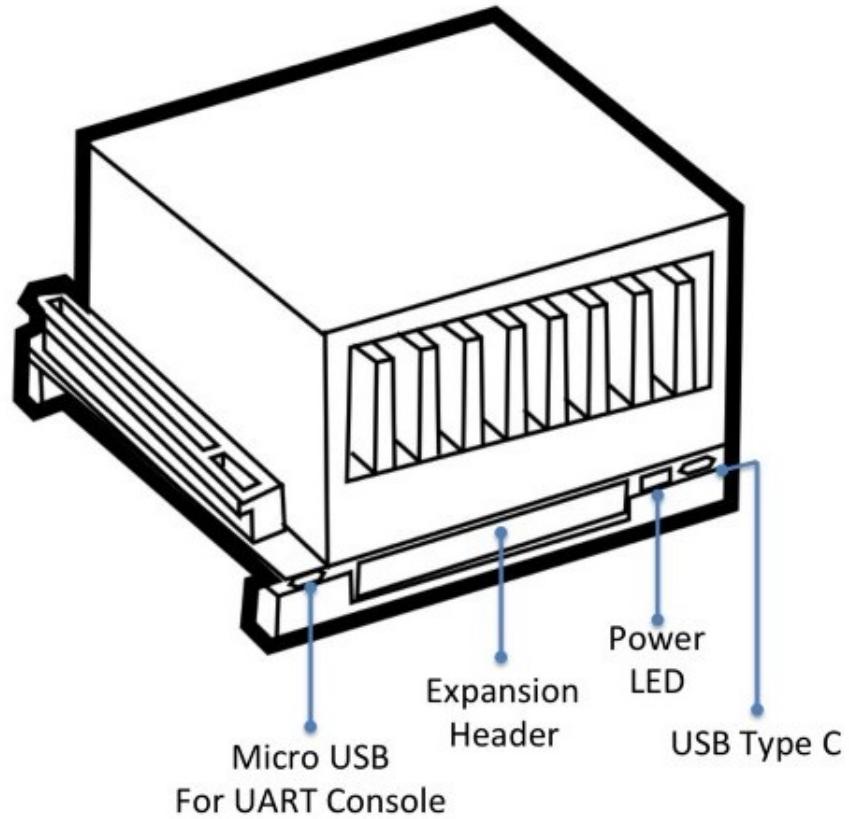
04. AI BOX 小試身手



01. Xavier 介紹



Xavier介紹





Xavier介紹



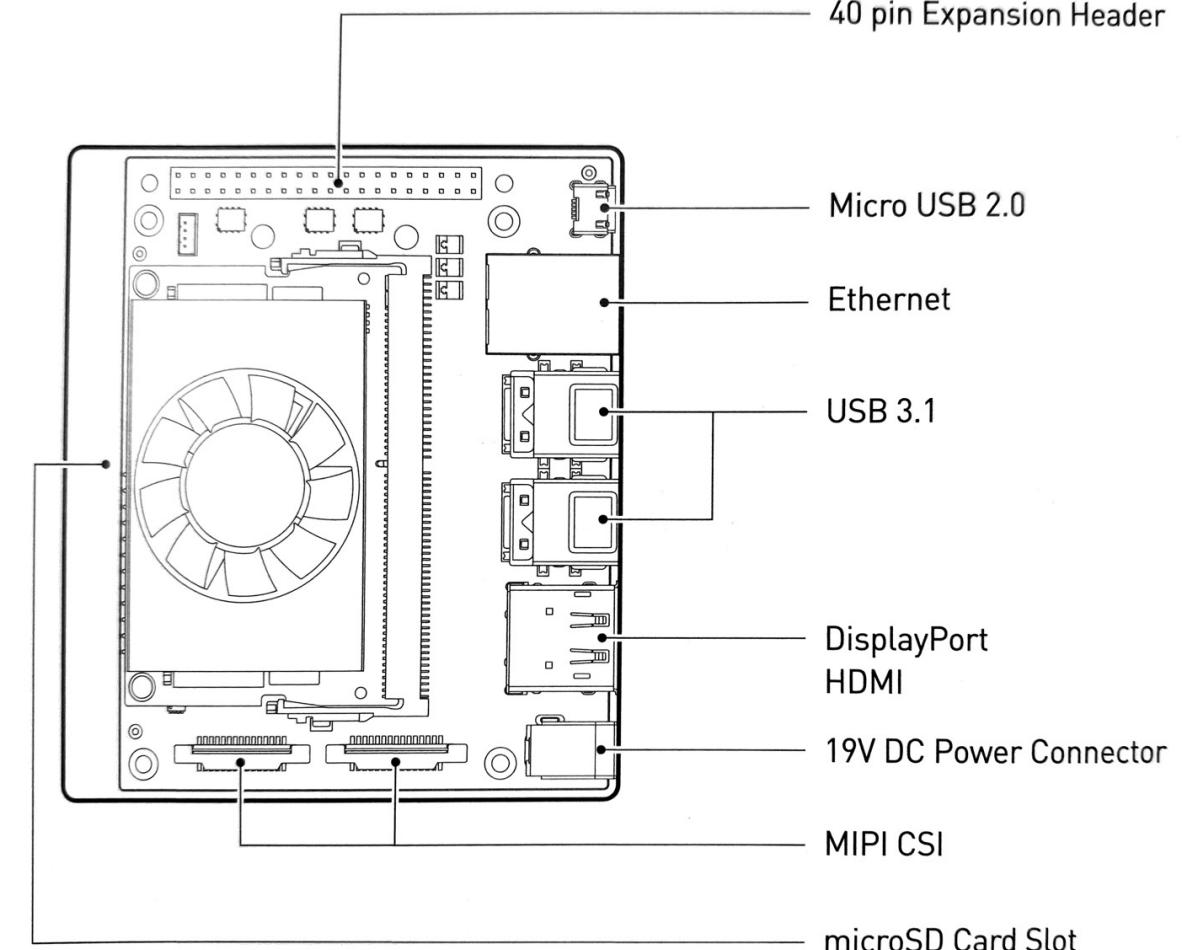
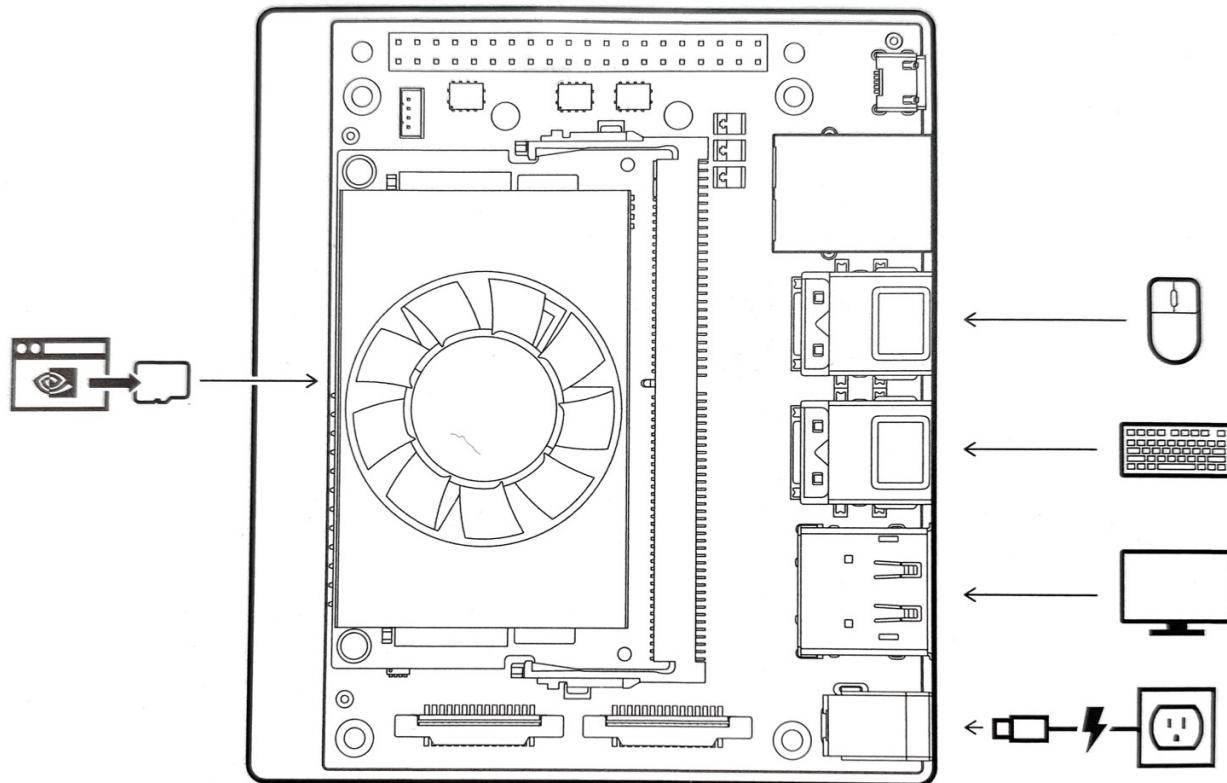
- 8核CPU
- 16G記憶體(32G)
- 32G儲存空間
- 體積小
- 省電
- 容易維護容易擴大規模
- 可適應較嚴苛的工廠環境
- DeepStream SDK
- 同樣算力下可使用到的RAM較大



02. Xavier NX 介紹



Xavier NX介紹



Xavier NX介紹

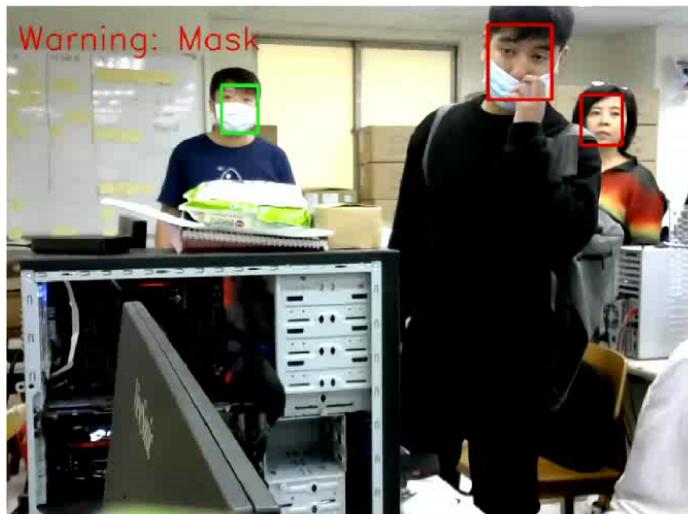
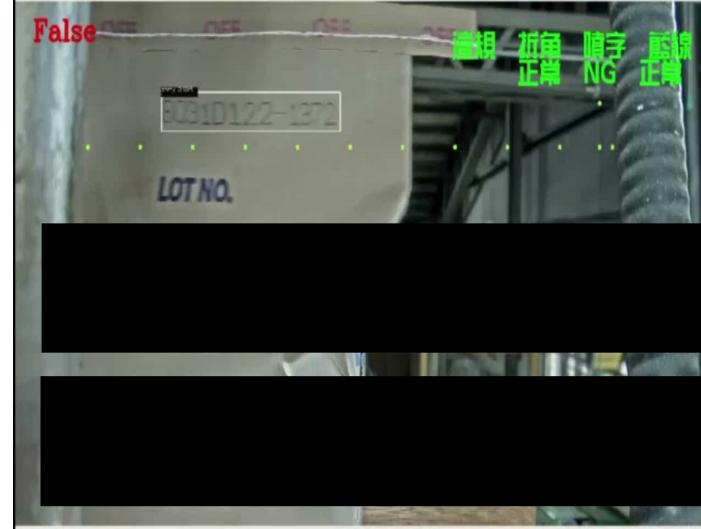
- 6核CPU
- 8G記憶體
- 沒有記憶體儲存空間，但可透過SD卡擴充
- 體積比xavier更小，效能為xavier的一半
- 省電
- 容易維護容易擴大規模
- 可適應較嚴苛的工廠環境
- DeepStream SDK
- 同樣算力下可使用到的RAM較大



03. 場景應用



場景應用





場景應用

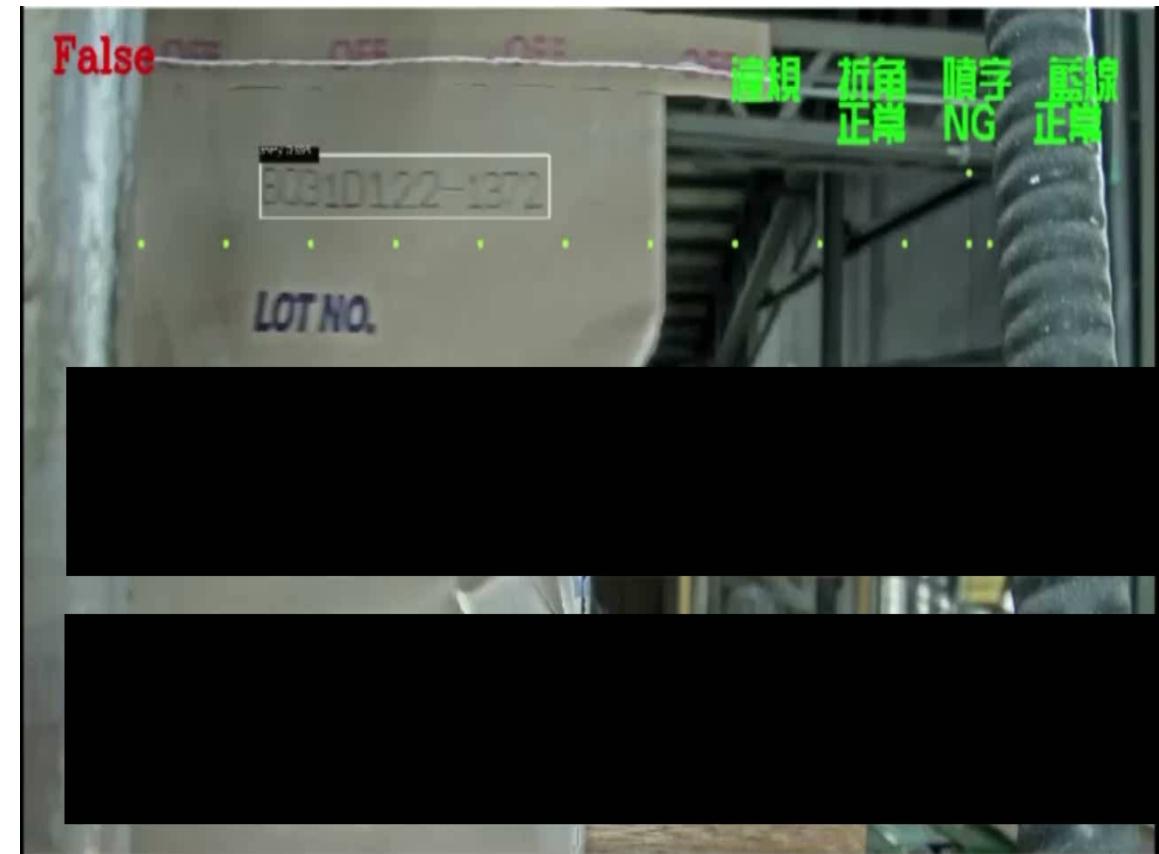
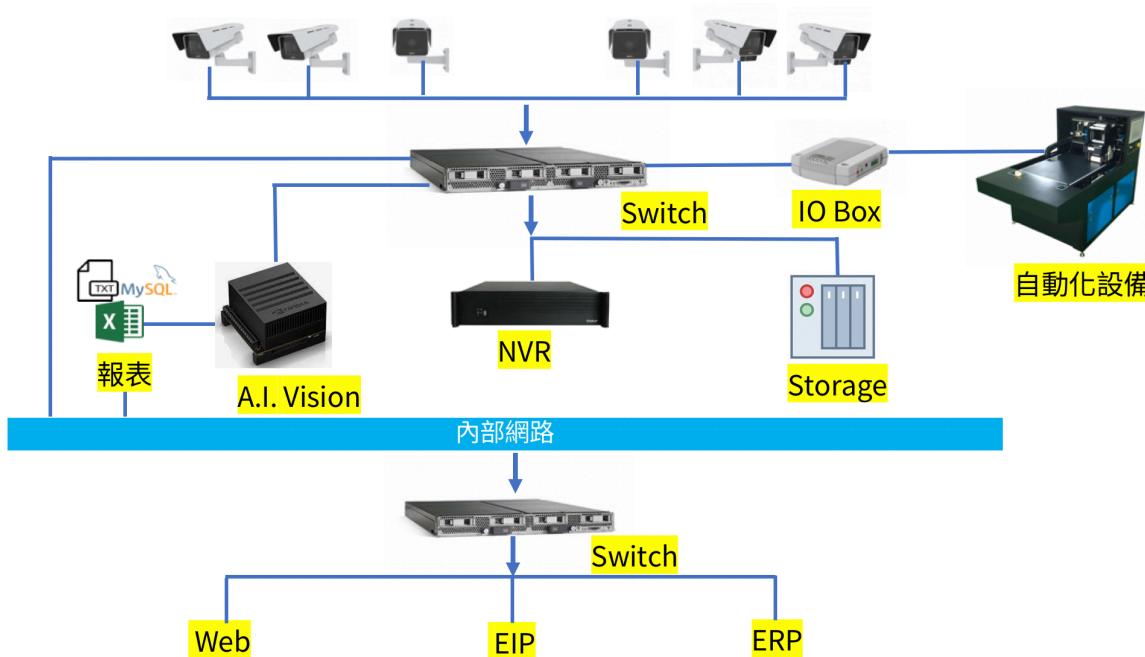
- 道路贓車、通緝車輛辨識
- 克服車上無法放置大型電腦





場景應用

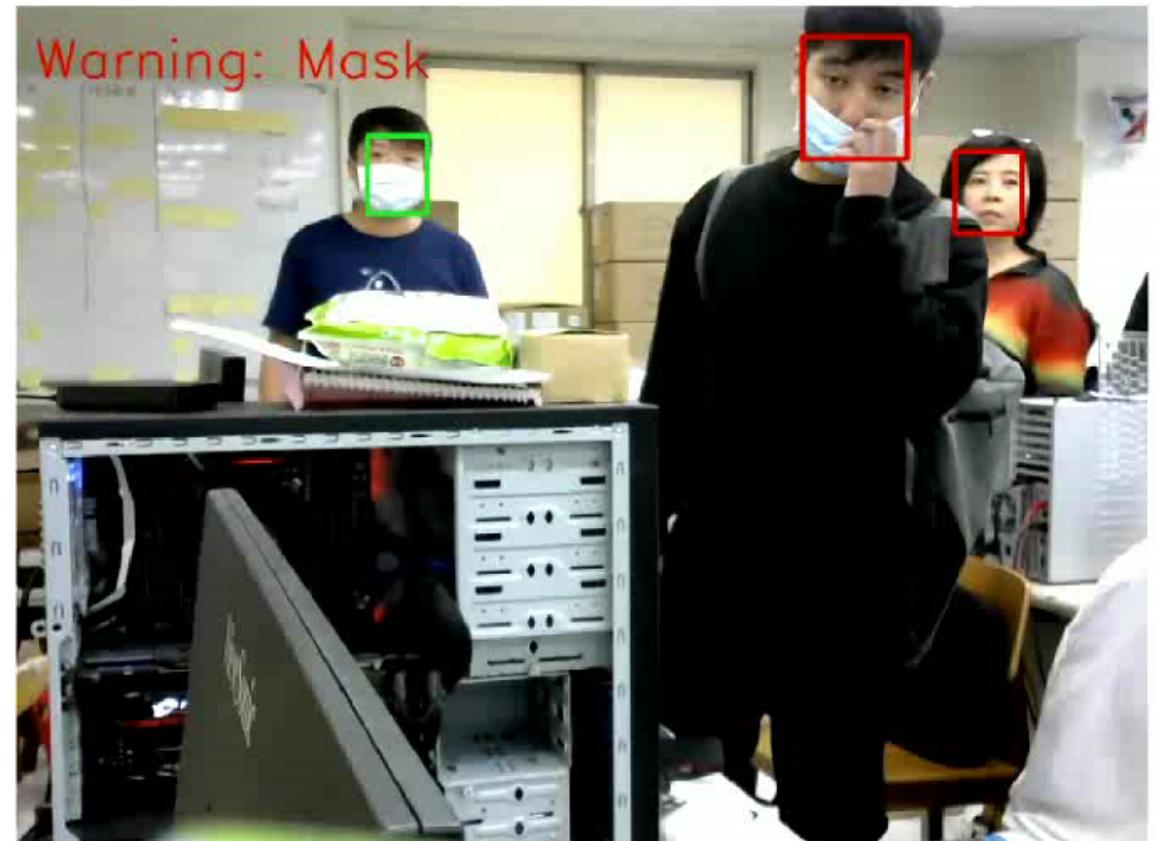
- 辨識不合格的包裝，發出警告後，使用機器手臂推掉
- 協助員工檢查包裝縫線





場景應用

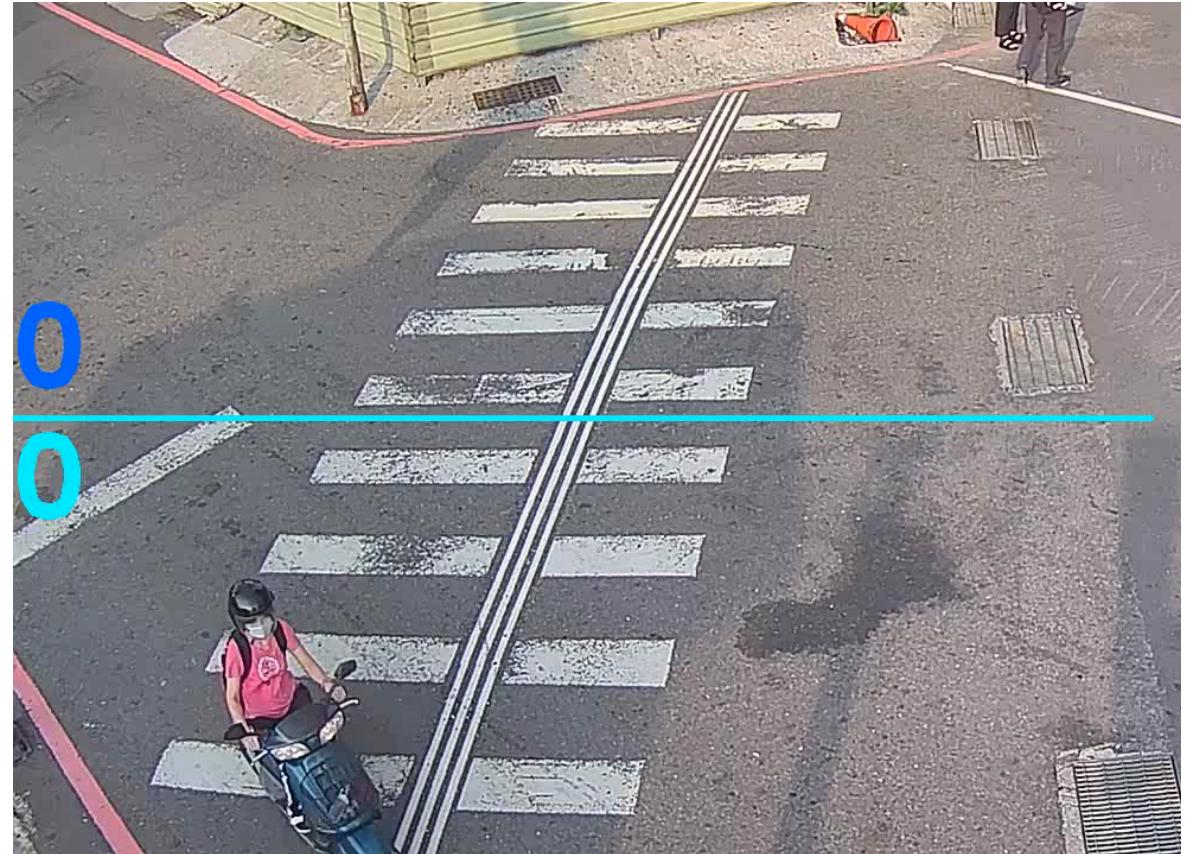
- 運用在大眾運輸、人多的公共場合
- 可搭配蜂鳴器





場景應用

- 計算人流
- 行人違規穿越馬路





04. AI BOX小試身手



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See below for downloadable documentation, software, and other resources.

JetPack 4.5.1 is available now! There are two main installation methods, depending on your developer kit:

SD Card Image Method

[JETSON XAVIER NX DEVELOPER KIT >](#)

[Download SD Card Image](#)

Follow the steps at [Getting Started with Jetson Xavier NX Developer Kit](#).

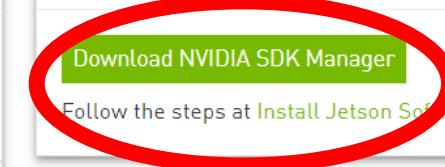
[JETSON NANO DEVELOPER KITS >](#)

NVIDIA SDK Manager Method

[FOR ANY JETSON DEVELOPER KIT >](#)

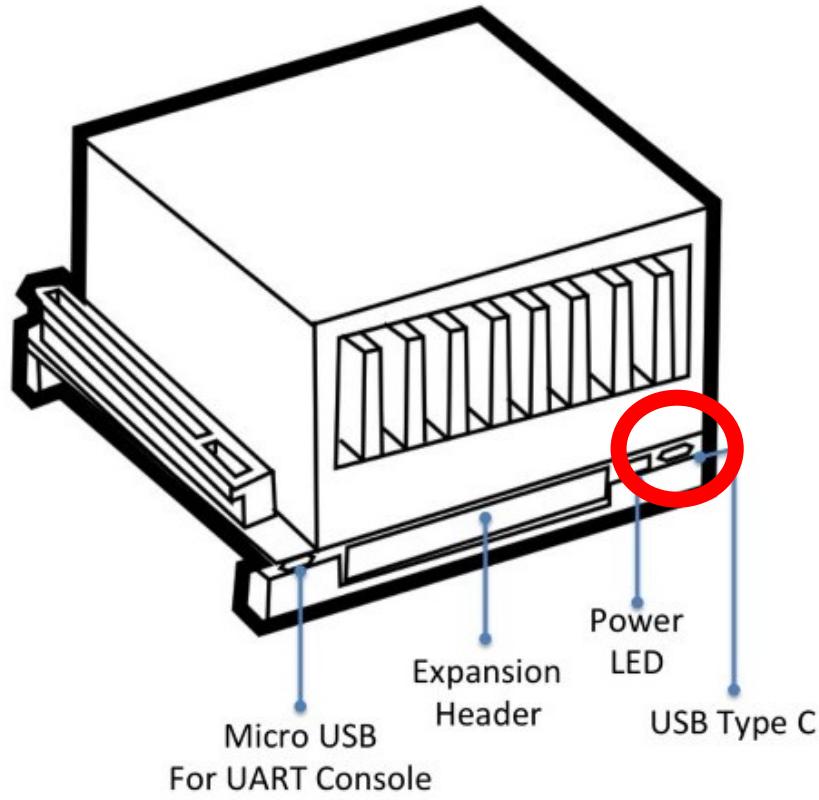
[Download NVIDIA SDK Manager](#)

Follow the steps at [Install Jetson Software with SDK Manager](#).

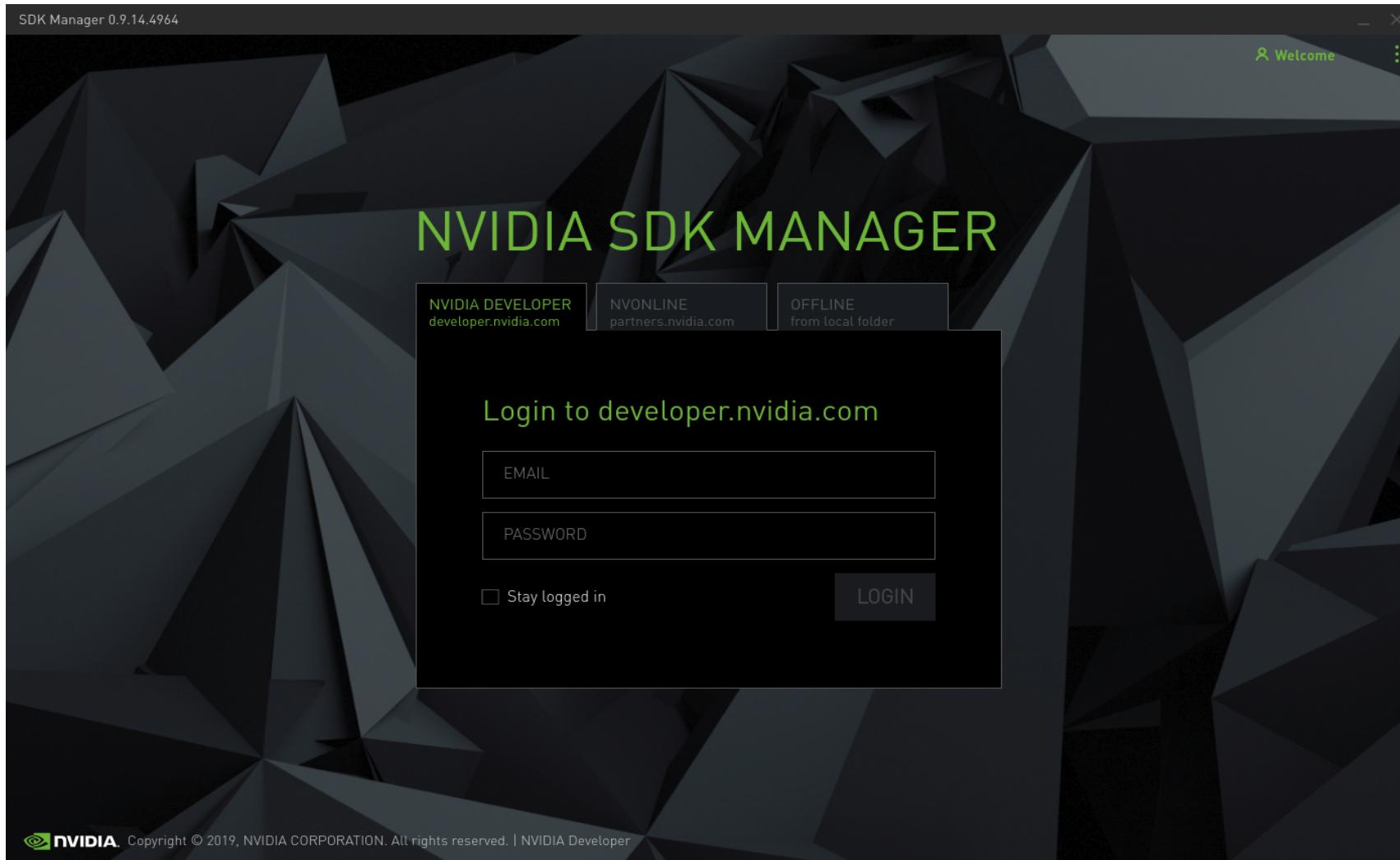




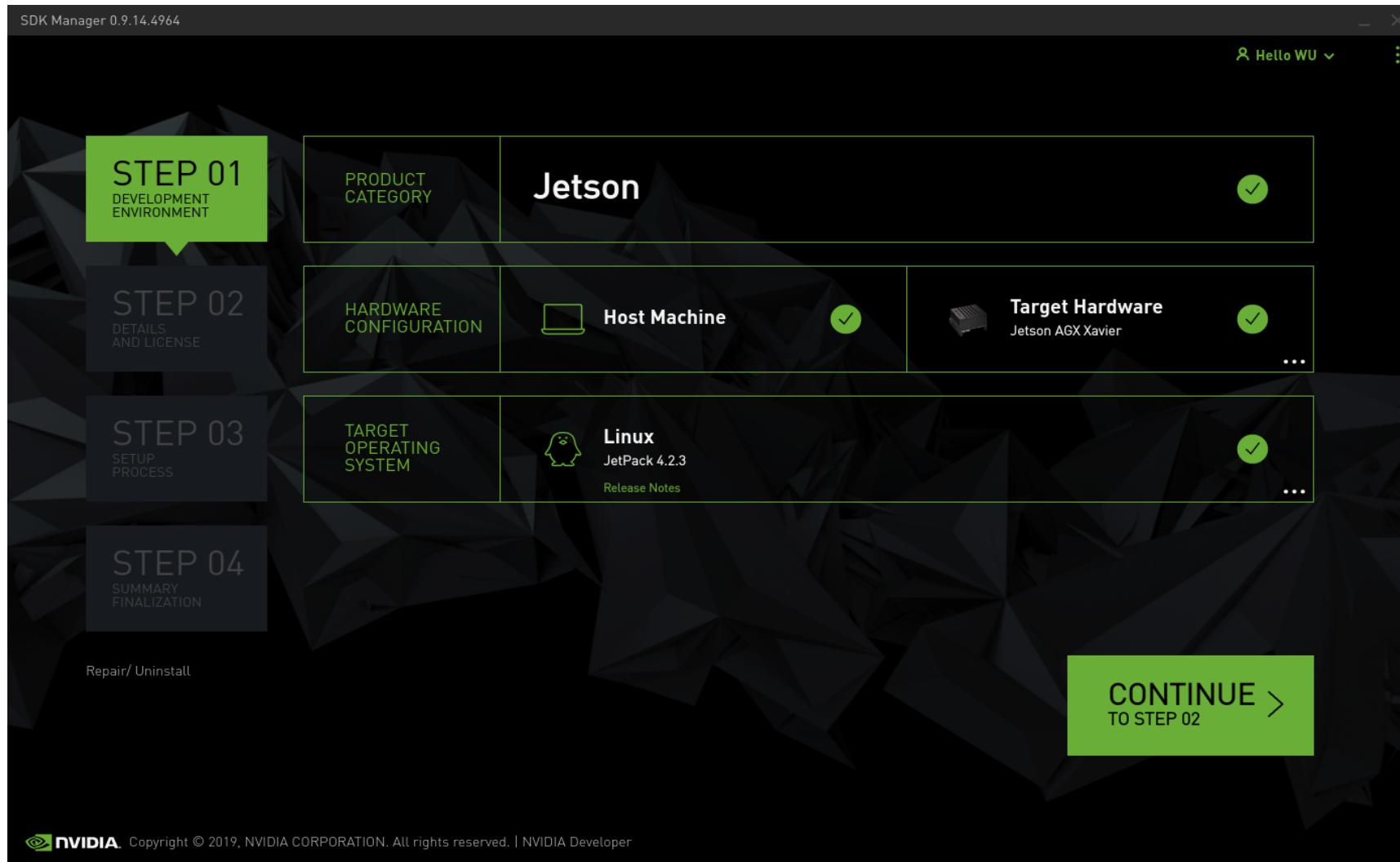
Xavier 刷機



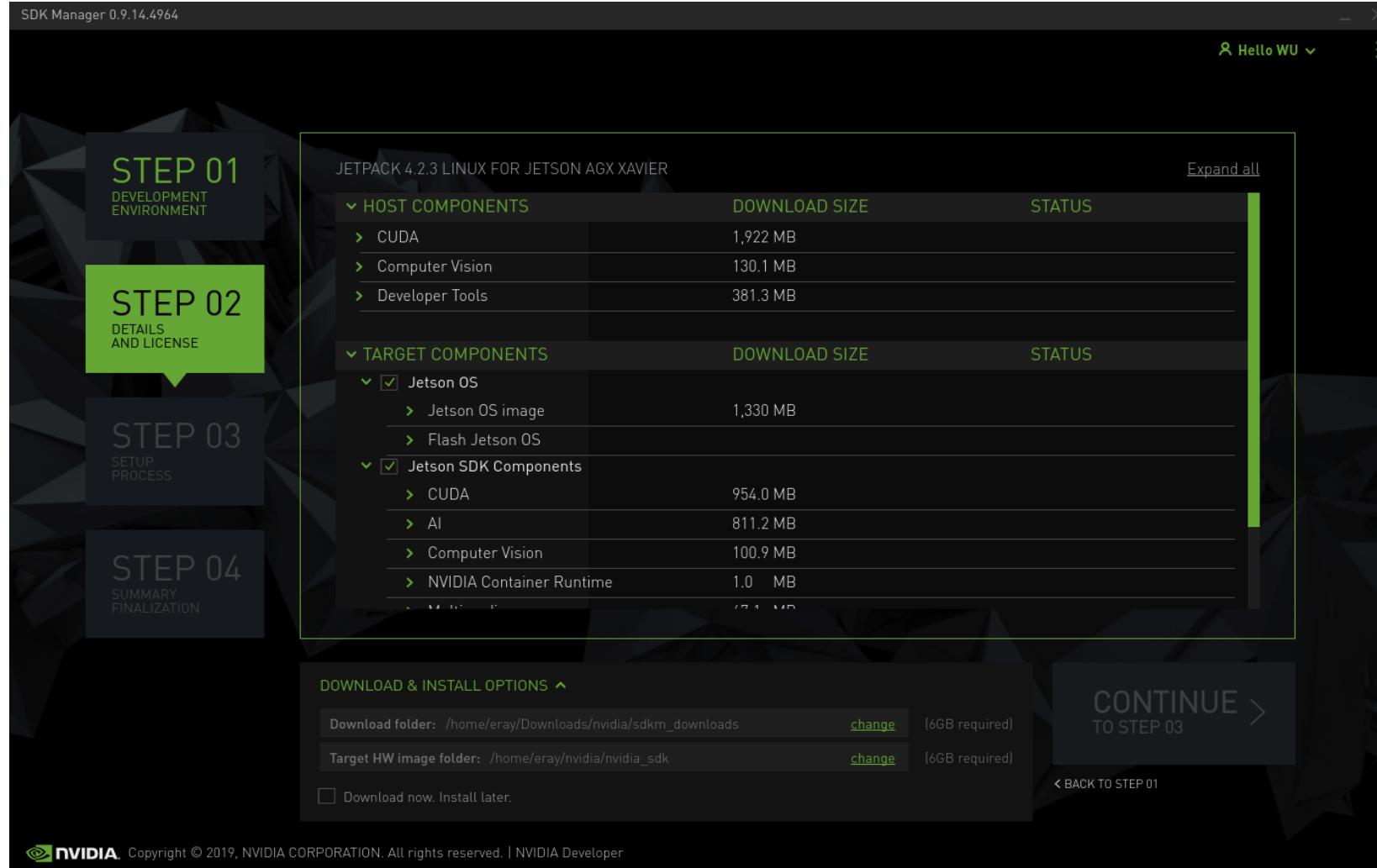
O · Xavier 刷機



Xavier 刷機



Xavier 刷機



Xavier 刷機



SDK Manager 0.9.14.4964

Hello WU

STEP 01
DEVELOPMENT ENVIRONMENT

STEP 02
DETAILS AND LICENSE

STEP 03
SETUP PROCESS

STEP 04
SUMMARY FINALIZATION

PAUSE FOR A BIT

DETAILS TERMINAL

JETPACK 4.2.3 LINUX FOR JETSON AGX XAVIER

Expand all

HOST COMPONENTS	DOWNLOAD SIZE	STATUS
CUDA	1,922 MB	Downloading - 3.9%
Computer Vision	130.1 MB	Pending download
Developer Tools	381.3 MB	Pending download

TARGET COMPONENTS	DOWNLOAD SIZE	STATUS
Jetson OS	1,330 MB	Downloading - 4.6%
Jetson OS image		Pending OS image
Flash Jetson OS		
Jetson SDK Components	954.0 MB	Downloading - 7.4%
CUDA	811.2 MB	Downloading - 15.4%
AI	100.9 MB	Downloading - 10.7%
Computer Vision		
NVIDIA Container Platform	1.0 MB	Pending download

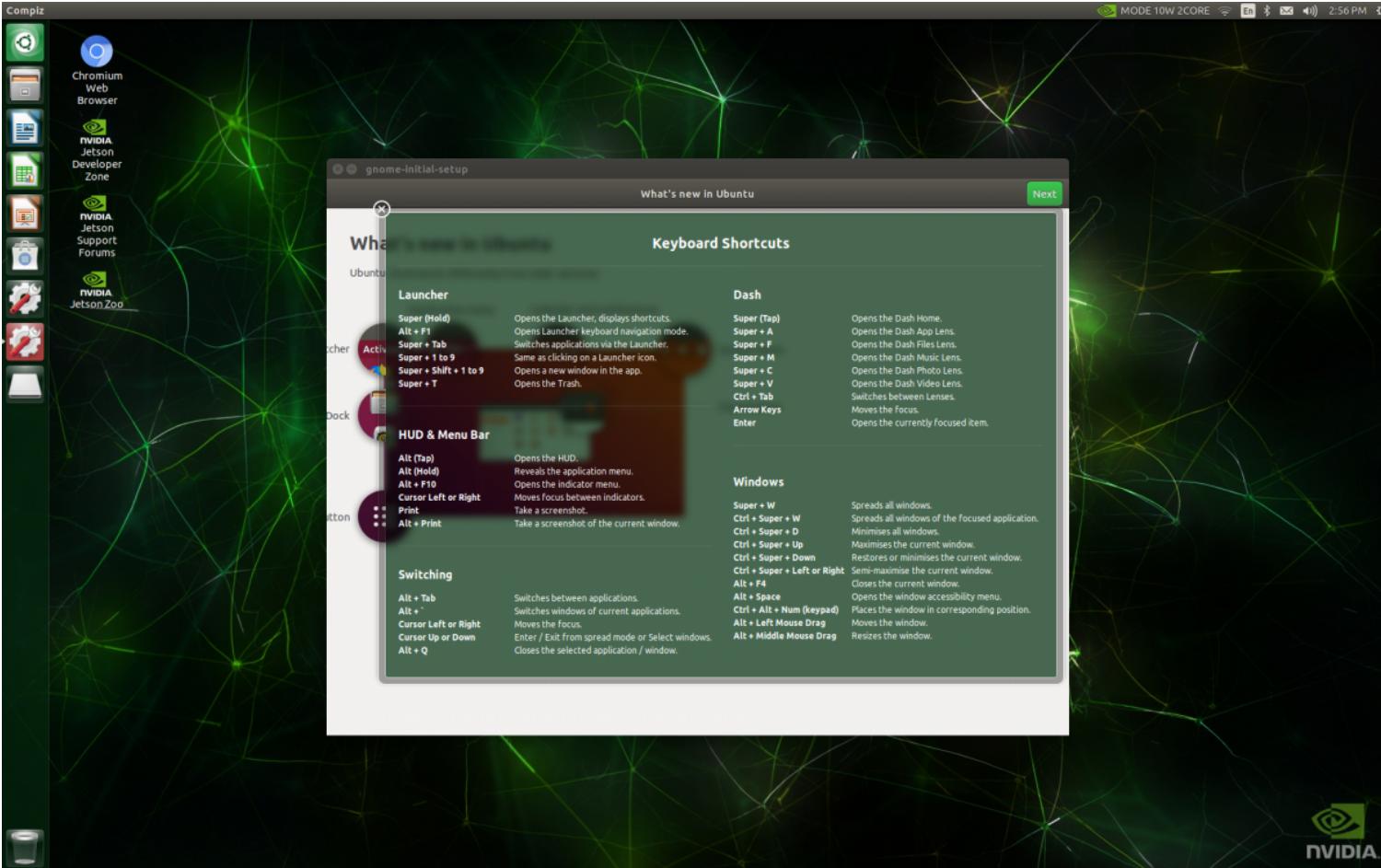
Download folder: /home/eray/Downloads/nvidia/sdkm_downloads

● Downloading: 5.79% (11.26MB/s)

● Installing: 0.00%

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Xavier 刷機



Xavier NX 刷機



A screenshot of the NVIDIA Developer website's header. It features the NVIDIA logo (a stylized eye icon) and the text "NVIDIA DEVELOPER". To the right are navigation links: HOME, BLOG, NEWS, FORUMS, DOCS, DOWNLOADS, and TRAINING.

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O· Xavier NX 刷機



Xavier NX 刷機



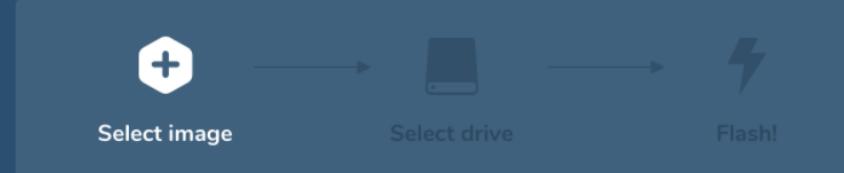
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 balenaEtcher

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Flash. Flawless.

Flash OS images to SD cards & USB drives, safely and easily.



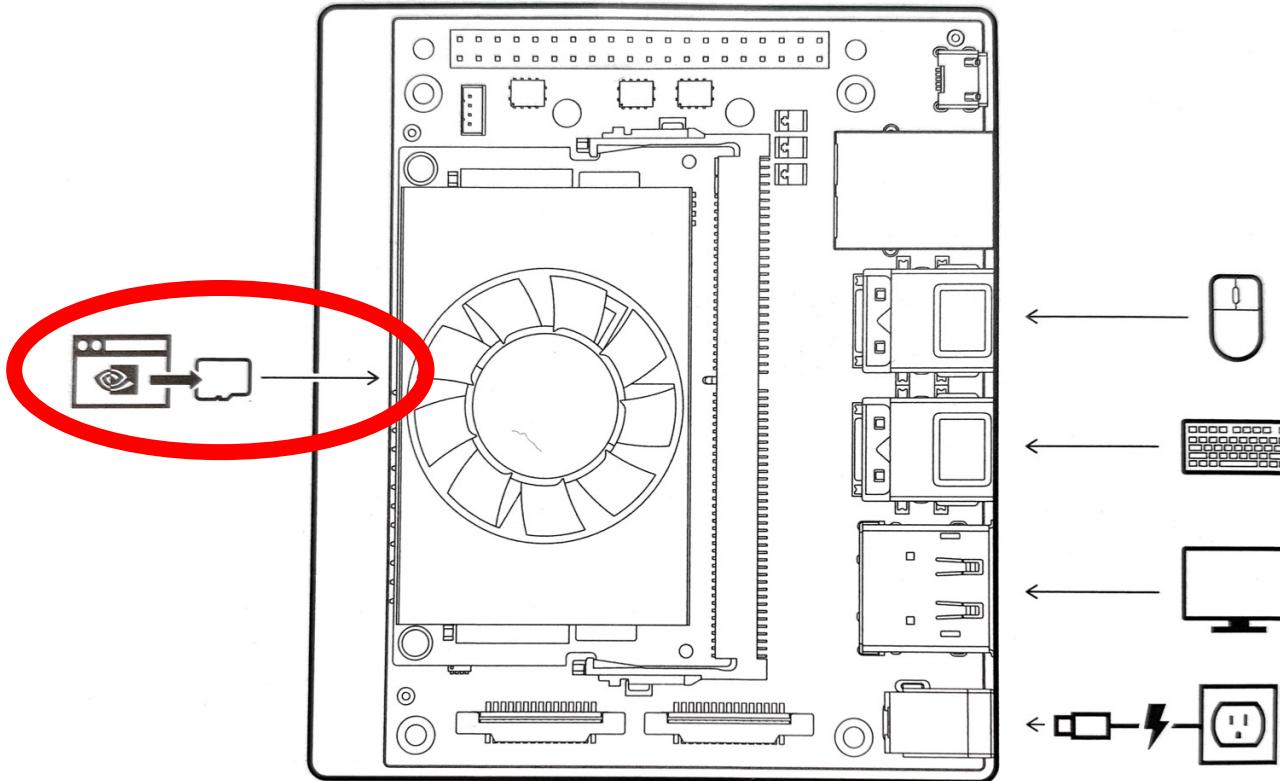
The diagram shows a three-step process: 1. A white hexagon icon with a plus sign, labeled "Select image". 2. An icon of a smartphone, labeled "Select drive". 3. A lightning bolt icon, labeled "Flash!". Arrows connect the icons from left to right.

Select image Select drive Flash!

[Download for Windows \(x86|x64\)](#)

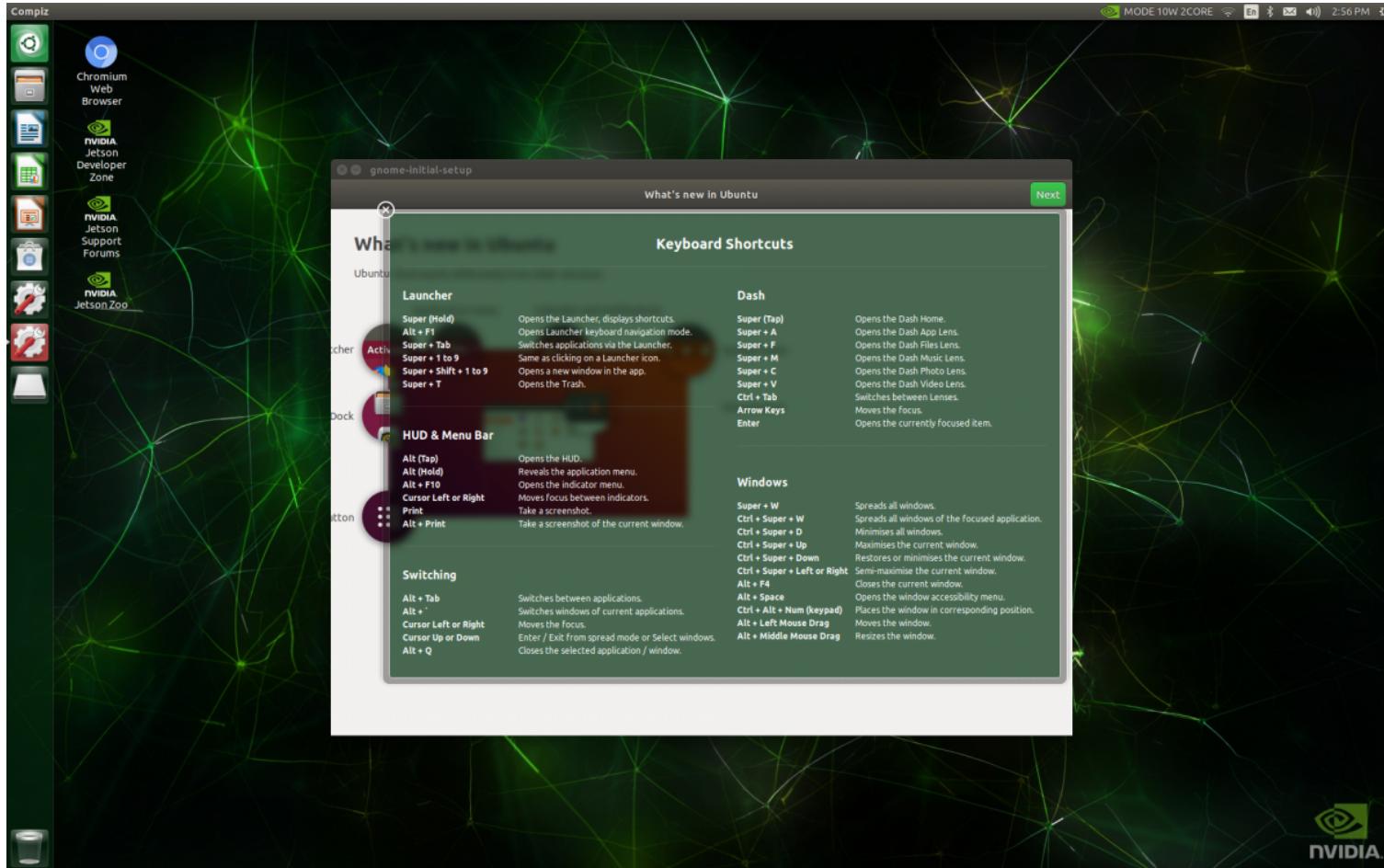
v1.5.116 [See what's new](#)

Xavier NX 刷機





Xavier NX 刷機





Xavier 系統相關指令



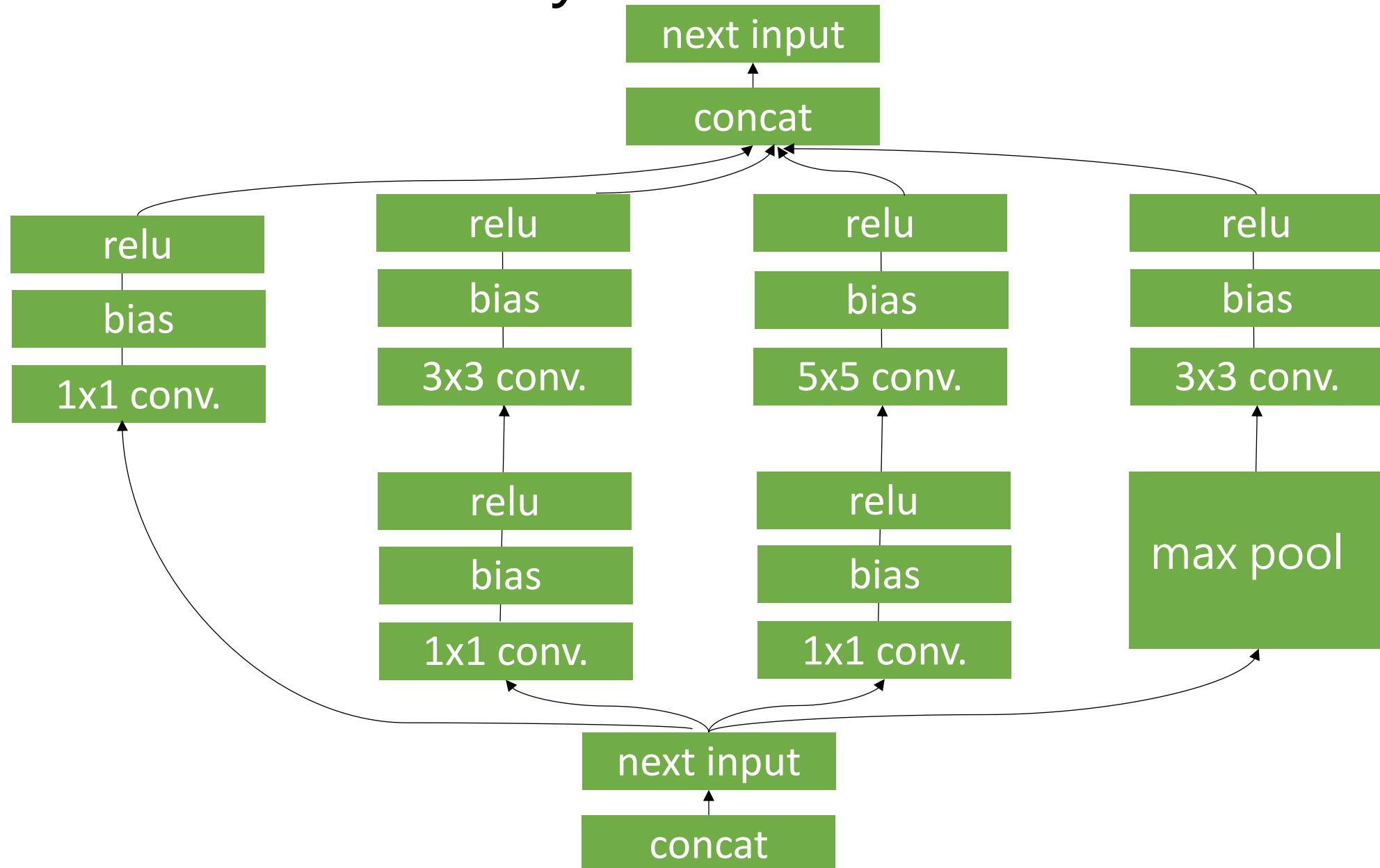
- 輸入 `sudo nvpmode1 -m 0` 將電力效能全開
- 輸入 `sudo jetson_clocks` 開啟風扇



TensorRT

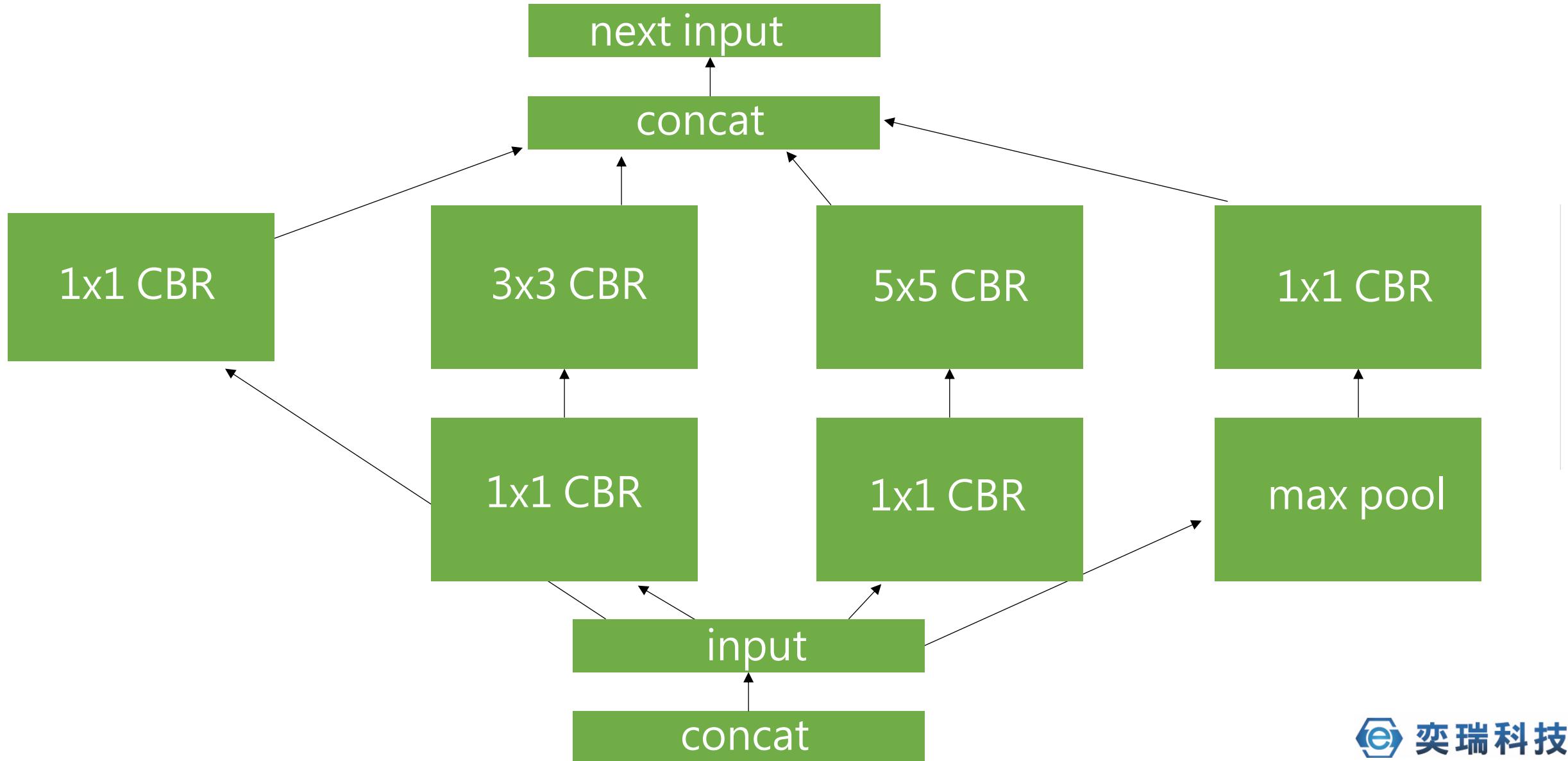
- 主要將部分layer進行合併，並降低精度以提升速度。
- 優點：
 - 提升辨識速度
 - 節省記憶體使用量
- 缺點：
 - 精確度稍微下降

Fuse network layers Inception structure in GoogleNet





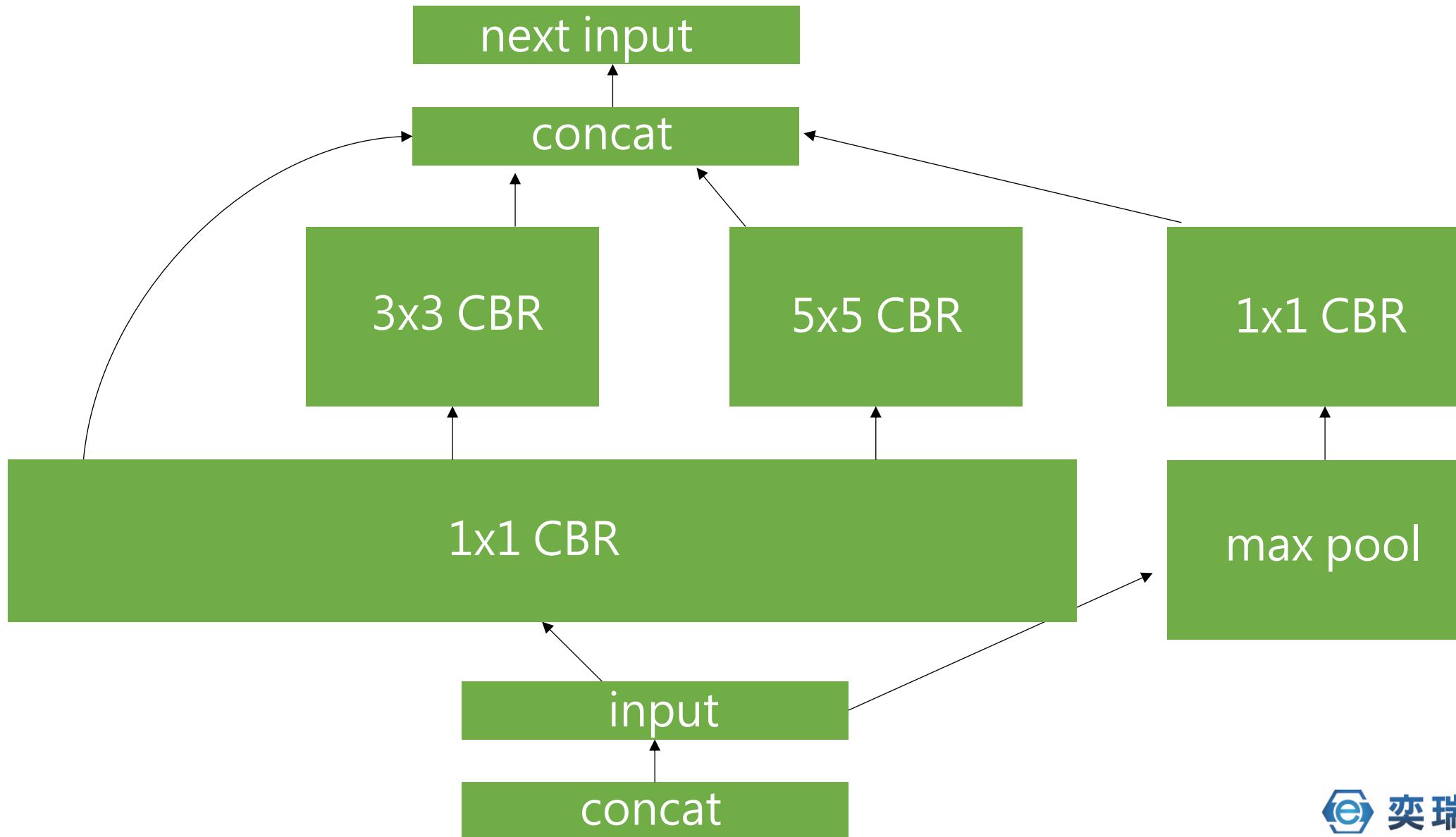
Fuse network layers Vertical fusion





Fuse network layers

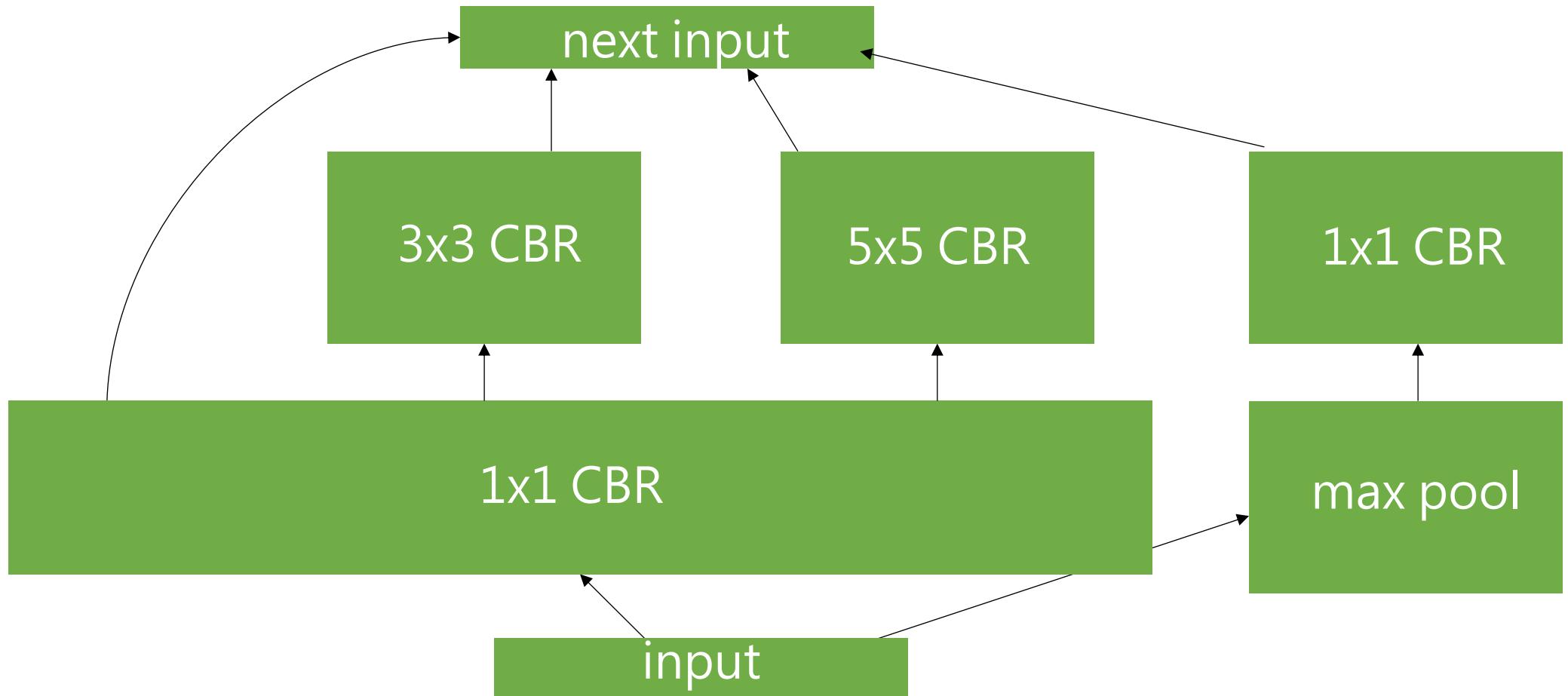
Horizontal fusion





Fuse network layers

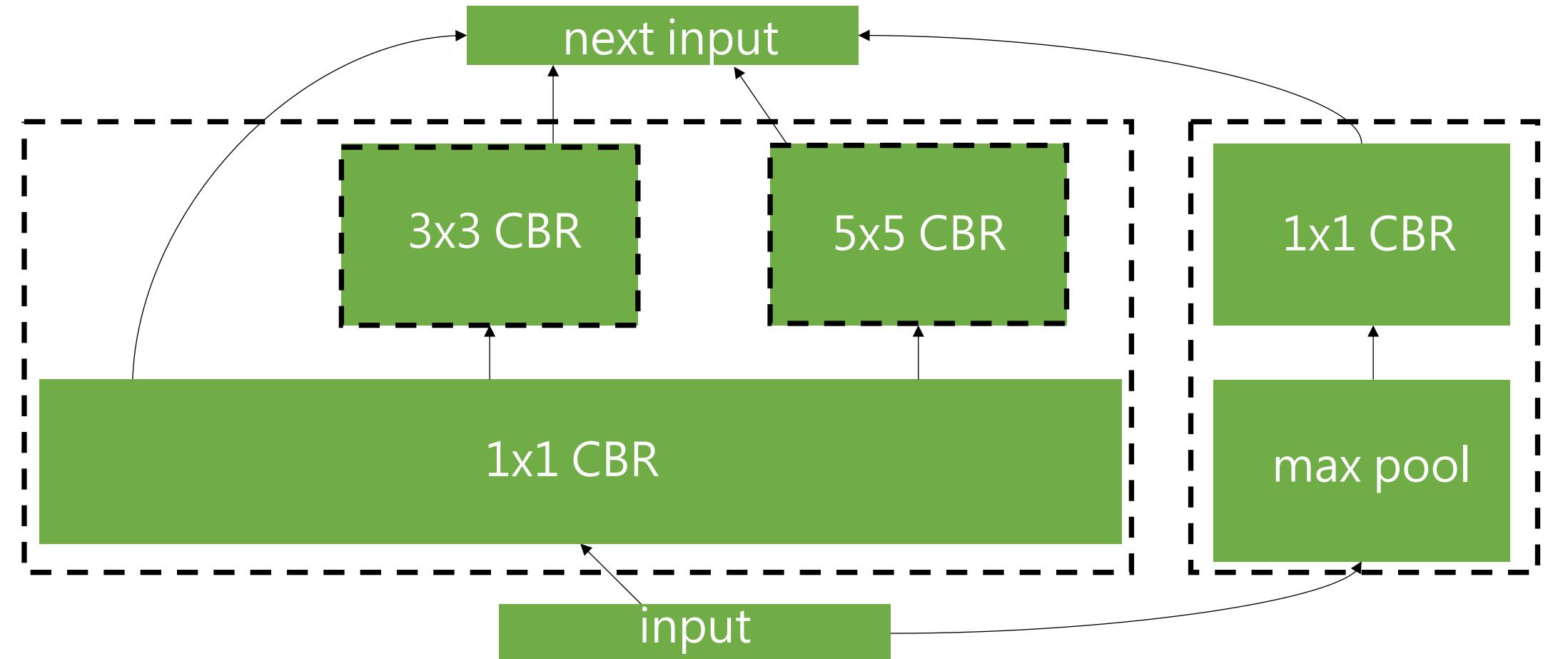
Concat elision





Fuse network layers

Concurrency





TensorRT 執行

- 1. 下載專案
- 2. 編譯以及安裝相依套件
- 3. 編譯tensorRT plugins(yolo layer)
- 4. 將model換成指定的檔案名稱
- 5. darknet model轉換成onnx
- 6. 將model轉換成TensorRT engine
- 7. TensorRT engine推論





下載專案

Project's github:

https://github.com/jkjung-avt/tensorrt_demos.git

Terminal:

```
$ git clone https://github.com/jkjung-avt/tensorrt_demos.git
```

The screenshot shows the GitHub repository page for 'jkjung-avt/tensorrt_demos'. The repository has 9 issues, 257 forks, and 786 stars. It contains code for TensorRT YOLOv4, YOLOv3, SSD, MTCNN, and GoogLeNet. The 'Code' tab is selected, showing the master branch. A 'Clone' button is available, along with a 'Download ZIP' button. The repository structure includes subfolders for common, doc, googlenet, mtcnn, plugins, ssd, utils, and yolo.

```
eray@eray-xavier:~$ git clone https://github.com/jkjung-avt/tensorrt_demos.git
Cloning into 'tensorrt_demos'...
remote: Enumerating objects: 94, done.
remote: Counting objects: 100% (94/94), done.
remote: Compressing objects: 100% (71/71), done.
remote: Total 1381 (delta 56), reused 49 (delta 23), pack-reused 1287
Receiving objects: 100% (1381/1381), 189.84 MiB | 8.34 MiB/s, done.
Resolving deltas: 100% (877/877), done.
Checking out files: 100% (85/85), done.
eray@eray-xavier:~$
```



編譯以及安裝相依套件



1. Install protobuf 3.8.0

```
$ wget https://raw.githubusercontent.com/jkjung-avt/jetson_nano/master/install_protobuf-3.8.0.sh
```

```
$ chmod +x ./install_protobuf-3.8.0.sh; ./install_protobuf-3.8.0.sh
```

2. Install pycuda

```
$ cd ${HOME}/project/tensorrt_demos/ssd
```

```
$ ./install_pycuda.sh
```

3. Install onnx

```
$ sudo pip3 install onnx==1.4.1
```



編譯tensorRT plugins(yolo layer)

```
$ cd ./tensorrt_demos/plugins
```

```
$ make all -j
```

The screenshot shows a terminal window with a dark background and light-colored text. The user is in the directory `~/tensorrt_demos/plugins`. They run `ls` to list files: `Makefile`, `README.md`, `yolo_layer.cu`, and `yolo_layer.h`. Then they run `make all -j`. The output shows the compilation process using `nvcc` and `g++` to generate `yolo_layer.o` from `yolo_layer.cu`, and `libyolo_layer.so` from `yolo_layer.o`. The final output is a shared library `libyolo_layer.so` and header files `yolo_layer.h` and `Makefile`.

```
eray@eray-xavier:~/tensorrt_demos/plugins
eray@eray-xavier:~/tensorrt_demos/plugins$ ls
Makefile README.md yolo_layer.cu yolo_layer.h
eray@eray-xavier:~/tensorrt_demos/plugins$ make all -j
nvcc -ccbin g++ -I"/usr/local/cuda/include" -I"/usr/local/TensorRT-7.1.3.4/include" -I"/usr/local/include" -I"plugin" -Xcompiler -fPIC -c -o yolo_layer.o yolo_layer.cu
g++ -shared -o libyolo_layer.so yolo_layer.o -L"/usr/local/cuda/lib64" -L"/usr/local/TensorRT-7.1.3.4/lib" -L"/usr/local/lib" -Wl,--start-group -lnvinfer -lnvplers -lnvinfer_plugin -lcudnn -lcublas -lcudart_static -lnvToolsExt -lcudart -lrt -ldl -lpthread -Wl,--end-group
eray@eray-xavier:~/tensorrt_demos/plugins$ ls
libyolo_layer.so README.md yolo_layer.h
Makefile yolo_layer.cu yolo_layer.o
eray@eray-xavier:~/tensorrt_demos/plugins$
```



將model轉換成TensorRT engine

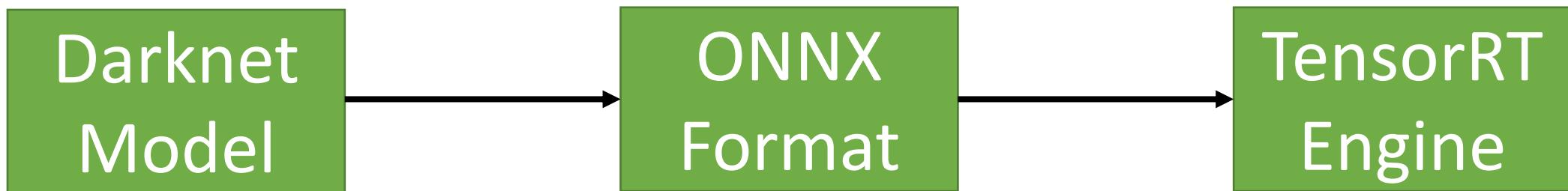
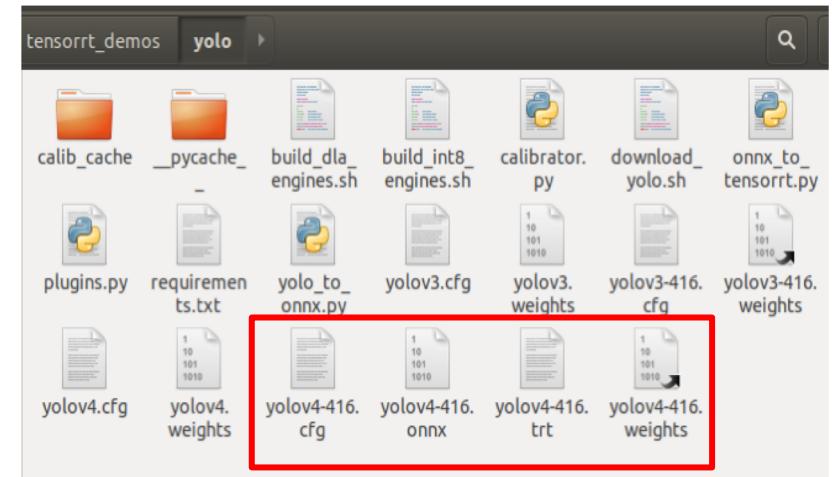


1. Prepare yolo models

2. Convert yolo models to tensorRT engines

```
$ python3 yolo_to_onnx.py -m yolov4-416
```

```
$ python3 onnx_to_tensorrt.py -m yolov4-416
```



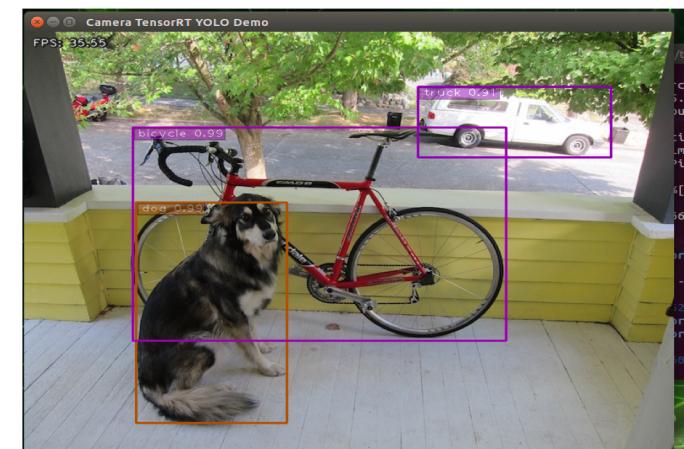


TensorRT engine推論

- 1.open your project directory
- 2.Prepare image
- 3.Open terminal and Detect image

```
$ python3 trt_yolo.py --image ${HOME}/Pictures/dog.jpg \ -m yolov4-416
```

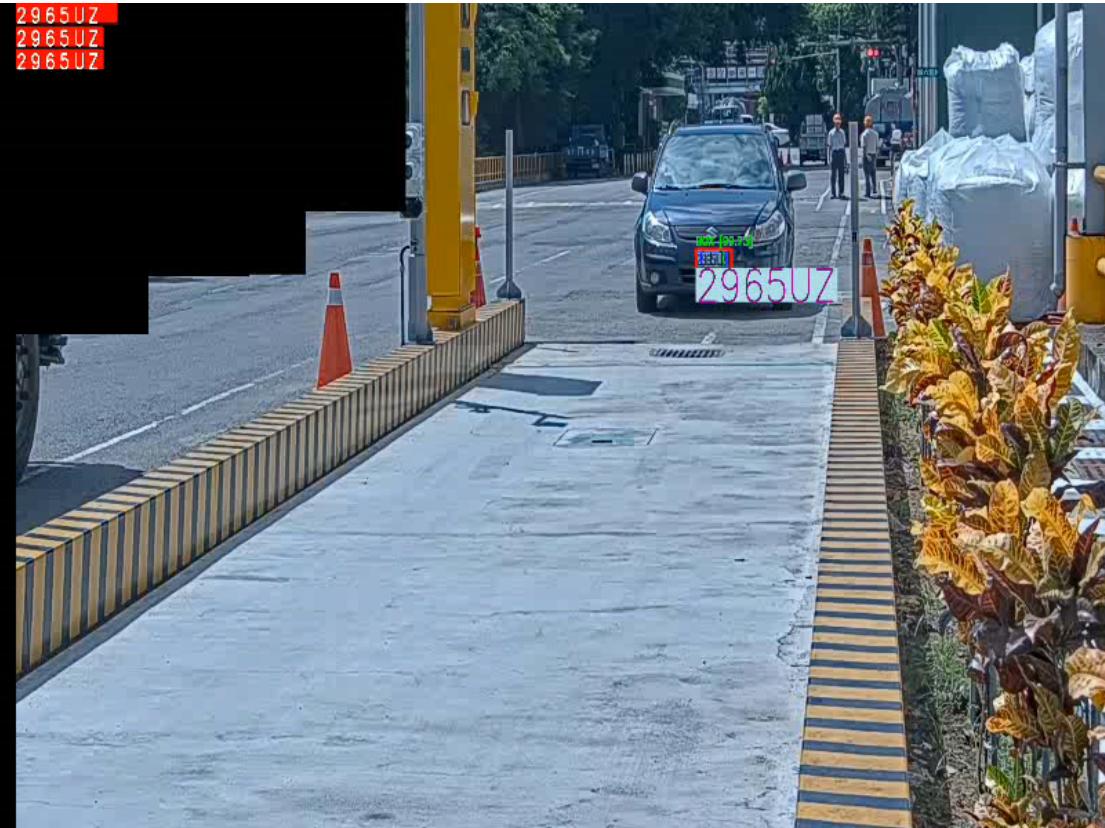
p.s **tensorRT engine不可以跨平台使用**，每個編譯出來的**tensorRT engine**都會根據平台進行最佳化的動作





Xavier未使用TensorRT

FPS: 0.76





Xavier使用TensorRT



FPS: 11.50





Xavier使用TensorRT差異



未使用TensorRT

使用TensorRT

20 FPS

36-42 FPS



DeepStream

- DeepStream是一套NVIDIA針對Video處理的一個FrameWork，它可以對多個輸入源的解碼、推論、顯示進行非同步及平行處理。
優點：

1. 使用者只需要專心撰寫推論部份即可。
2. 所有動作都在GPU上運作，省下CPU傳輸至GPU的時間。
3. 內建在GPU上運行的追蹤演算法，節省偵測時間。
4. 透過非同步、及平行處理加速運算效能。

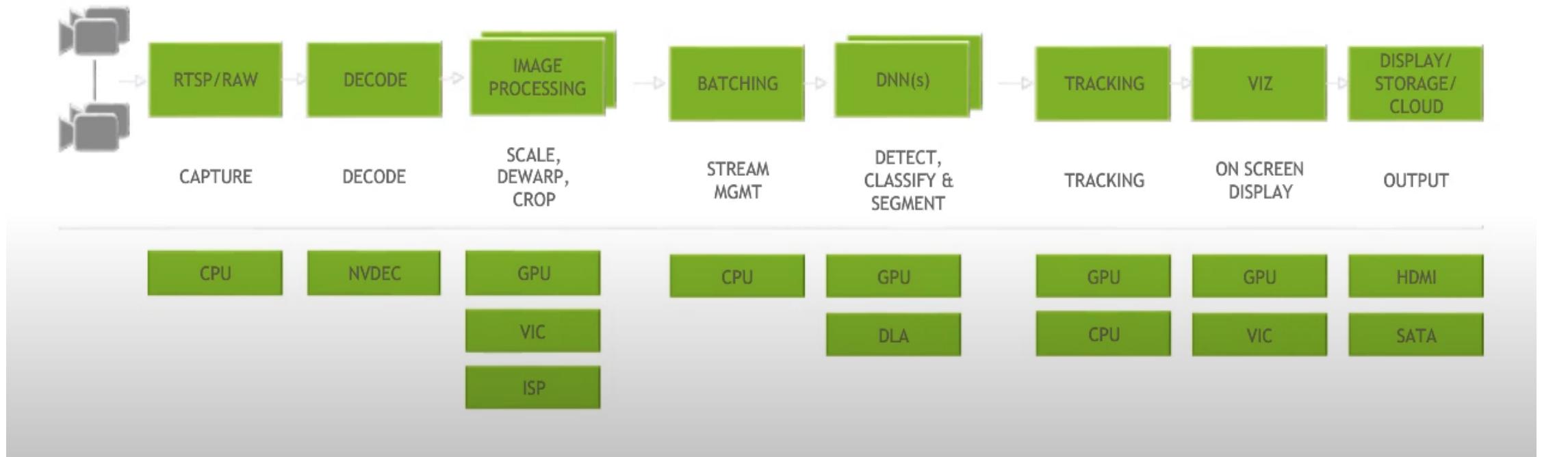
- 缺點：

1. 若要做較精細的處理，需要了解GStreamer的運作以及撰寫其plugins





DEEPSTREAM GRAPH ARCHITECTURE





Deepstream 實作 YOLOV3 Model



Step



1. 安裝deepStream
2. 編輯檔案(更改class_count, anchors) and 編譯
3. 在要運行的config中指定已經編譯完成的.so的路徑運行程式
4. 開啟python檔案進行設定，

編輯檔案(更改class_count,anchors)and 編譯-1

- Enter to the folder:
- \$ cd /opt/nvidia/deepstream/deepstream/sources/objectDetector_Yolo
- Download model and config
- \$ sudo chmod+x prebuild.sh; sudo ./prebuild.sh
- Edit the yolo layer file(class count, anchors)
- \$ sudo vim nvdsinfer_custom_impl_Yolo/nvdsparsebbox_Yolo.cpp
- Compiler .so
- \$ cd nvdsinfer_custom_impl_Yolo; sudo make all

```
606  
607 [yolo]  
608 mask = 6,7,8  
609 anchors = 10,13, 16,30, 33,23, 30,61, 62,45, 59,119, 116,90, 156,198, 373,326  
610 classes=80  
611 num=9  
612 jitter=.3  
613 ignore_thresh = .7
```

Download model and config
\$ sudo chmod+x prebuild.sh; sudo ./prebuild.sh

Edit the yolo layer file(class count, anchors)

```
33 static const int NUM_CLASSES_YOLO = 80;  
  
274 /* C-linkage to prevent name-mangling */  
275 extern "C" bool NvDsInferParseCustomYoloV3(  
276     std::vector<NvDsInferLayerInfo> const& outputLayersInfo,  
277     NvDsInferNetworkInfo const& networkInfo,  
278     NvDsInferParseDetectionParams const& detectionParams,  
279     std::vector<NvDsInferParseObjectInfo>& objectList)  
280 {  
281     static const std::vector<float> kANCHORS = {  
282         10.0, 13.0, 16.0, 30.0, 33.0, 23.0, 30.0, 61.0,  
283         62.0, 45.0, 59.0, 119.0, 116.0, 90.0, 156.0, 198.0, 373.0,  
284         326.0};  
285     static const std::vector<std::vector<int>> kMASKS = {  
286         {6, 7, 8},  
287         {3, 4, 5},  
288         {0, 1, 2}};  
289     return NvDsInferParseYoloV3 (  
290         outputLayersInfo, networkInfo, detectionParams,  
objectList,  
290         kANCHORS, kMASKS);  
291 }
```

● 編輯 PGIE(Primary GPU Inference Engines) config

- **Edit the pgie config**
- \$ vim config_infer_primary_yoloV3.txt
- **parameter should be changed**
- custom-network-config=
- model-file=
- labelfile-path= (network's class_name)
- num-detected-classes= network's class count
- custom-lib-path= Please enter the yolo layer library you build.

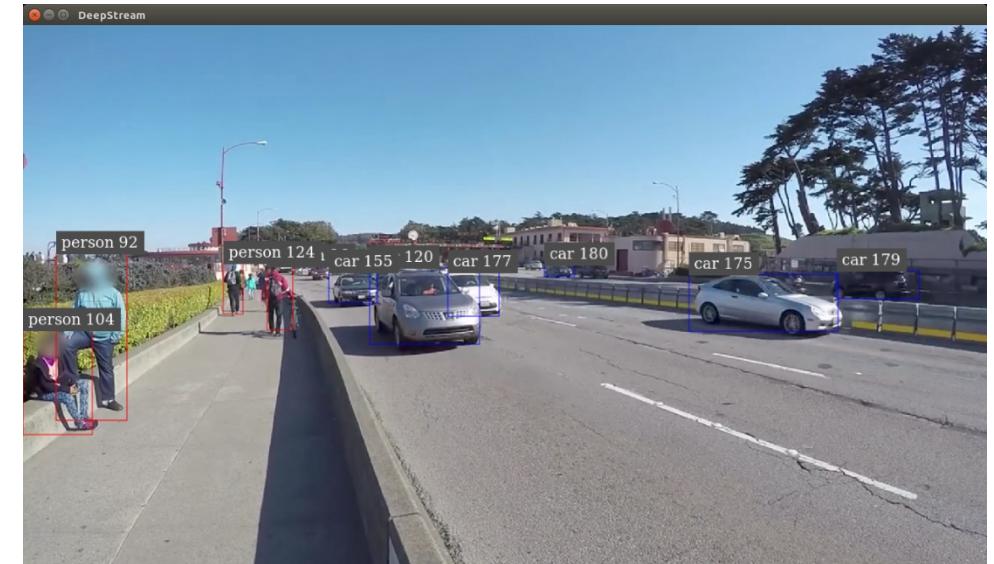
```
60 [property]
61 gpu-id=0
62 net-scale-factor=0.0039215697906911373
63 #0=RGB, 1=BGR
64 model-color-format=0
65 custom-network-config=yolov3.cfg
66 model-file=yolov3.weights
67 model-engine-file=yolov3_b1_gpu0_int8.engine
68 labelfile-path=labels.txt
69 int8-calib-file=yolov3-calibration.table.trt7.0
70 ## 0=FP32, 1=INT8, 2=FP16 mode
71 network-mode=1
72 num-detected-classes=80
73 gie-unique-id=1
74 network-type=0
75 is-classifier=0
76 ## 0=Group Rectangles, 1=DBSCAN, 2=NMS, 3= DBSCAN+NMS Hybrid, 4 = None(No clustering)
77 cluster-mode=2
78 maintain-aspect-ratio=1
79 parse-bbox-func-name=NvDsInferParseCustomYoloV3
80 custom-lib-path=nvdsinfer_custom_impl Yolo/libnvdsinfer_custom_impl Yolo.so
81 engine-create-func-name=NvDsInferYoloCudaEngineGet
82 #scaling-filter=0
83 #scaling-compute-hw=0
84
85 [class-attrs-all]
86 nms-iou-threshold=0.3
87 threshold=0.7
```



Test the projects



- **Enter to the folder:**
- \$ cd /opt/nvidia/deepstream/deepstream/sources/objectDetector_Yolo
- **run the scripts:**
- \$ deepstream-app -c deepstream_app_config_yoloV3.txt





Python run deepstream-1

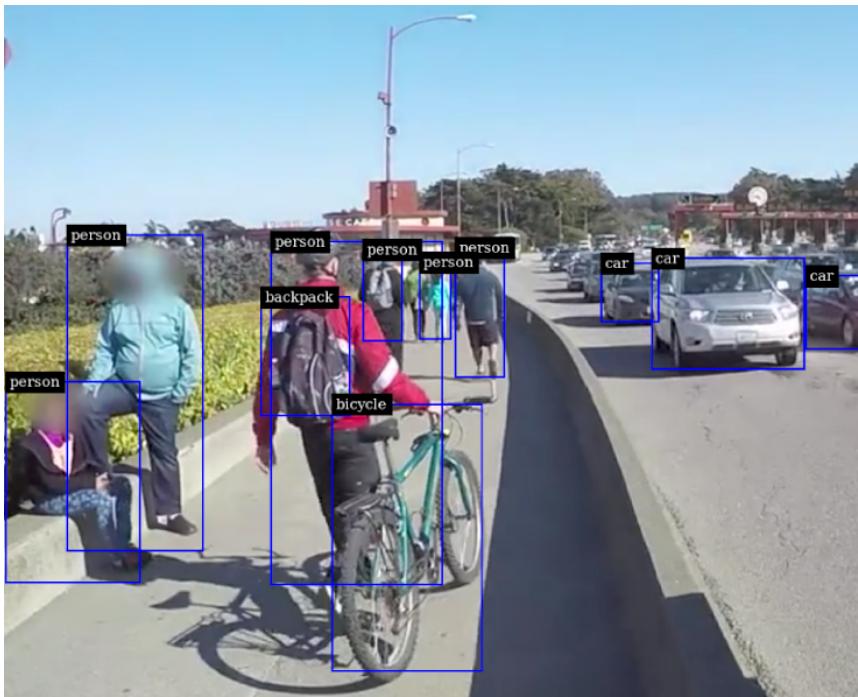


- **Clone deepstream-python example projects**
- \$ sudo git clone https://github.com/NVIDIA-AI-IOT/deepstream_python_apps
- **Copy apps/common and apps/deepstream-test1/deepstream_test_1.py to folder**
- \$ sudo copy -R deepstream_python_apps/apps/common ./; sudo copy deepstream_python_apps/apps/deepstream-test1/deepstream_test_1.py ./
- **Edit deepstream_test_1.py to load PGIE's config**
- \$sudo vim deepstream_test_1.py
- **Edit the line 204, change the config-file-path to your config_infer_primary_yoloV3.txt**

```
201     streammux.set_property('height', 1080)
202     streammux.set_property('batch-size', 1)
203     streammux.set_property('batched-push-timeout', 4000000)
204     pgie.set_property('config-file-path', "/opt/nvidia/deepstream/deepstream
      -5.0/sources/objectDetector_Yolo/config_infer_primary_yoloV3.txt")
```

Python run deepstream-2

- Run Code
- \$ sudo python3 deepstream_test_1.py ../../samples/streams/sample_720p.h264



```
eray@eray-xavier:/opt/nvidia/deepstream/deepstream/sources/objectDetector... x eray@eray-xavier:/opt/nvidia/deepstream/deepstream/sources/objectDetector...
41 def osd_sink_pad_buffer_probe(pad,info,u_data):
42     frame_number=0
43     #initializing object counter with 0.
44     obj_counter = {
45         PGIE_CLASS_ID_VEHICLE:0,
46         PGIE_CLASS_ID_PERSON:0,
47         PGIE_CLASS_ID_BICYCLE:0,
48         PGIE_CLASS_ID_ROADSIGN:0
49     }
50     num_rects=0
51     gst_buffer = info.get_buffer()
52     if not gst_buffer:
53         print("Unable to get GstBuffer ")
54         return
55
56     # Retrieve batch metadata from the gst buffer
57     # Note that pyds.gst_buffer_get_nvds_batch_meta() expects the
58     # C address of gst_buffer as input, which is obtained with hash(gst_buffer)
59     batch_meta = pyds.gst_buffer_get_nvds_batch_meta(hash(gst_buffer))
60     l_frame = batch_meta.frame_meta_list
61     while l_frame is not None:
62         try:
63             # Note that l_frame.data needs a cast to pyds.NvDsFrameMeta
64             # The casting is done by pyds.glist_get_nvds_frame_meta()
65             # The casting also keeps ownership of the underlying memory
66             # in the C code, so the Python garbage collector will leave
67             # it alone.
68             #frame_meta = pyds.glist_get_nvds_frame_meta(l_frame.data)
69             frame_meta = pyds.NvDsFrameMeta.cast(l_frame.data)
70             except StopIteration:
71                 break
72
73             frame_number=frame_meta.frame_num
74             num_rects = frame_meta.num_obj_meta
75             l_obj=frame_meta.obj_meta_list
76             while l_obj is not None:
77                 try:
78                     # Casting l_obj.data to pyds.NvDsObjectMeta
79                     #obj_meta=pyds.glist_get_nvds_object_meta(l_obj.data)
80                     obj_meta=pyds.NvDsObjectMeta.cast(l_obj.data)
81                     except StopIteration:
82                         break
83                     if obj_meta.class_id < 4:
84                         obj_counter[obj_meta.class_id] += 1
85                     obj_meta.rect_params.border_color.set(0.0, 0.0, 1.0, 0.0)
86                     try:
87                         l_obj=l_obj.next
88                     except StopIteration:
89                         break
90
91                     # Acquiring a display meta object. The memory ownership remains in
92                     # the C code so downstream plugins can still access it. Otherwise
93                     # the garbage collector will claim it when this probe function exits.
94                     #display_meta=pyds.nvds_acquire_display_meta_from_pool(batch_meta)
95                     display_meta.num_labels = 1
96
97
98
99
99
```

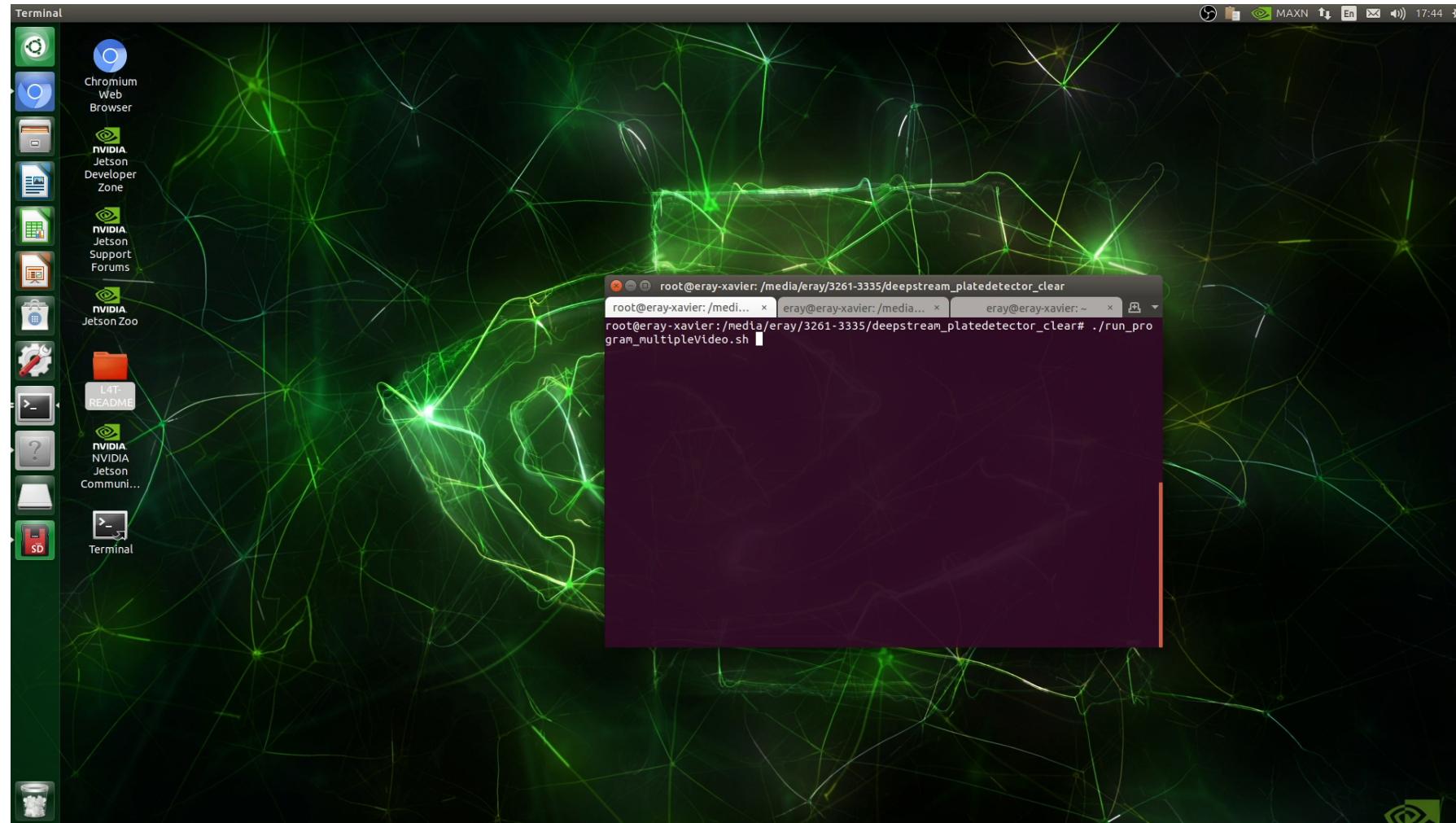


Xavier未使用DeepStream



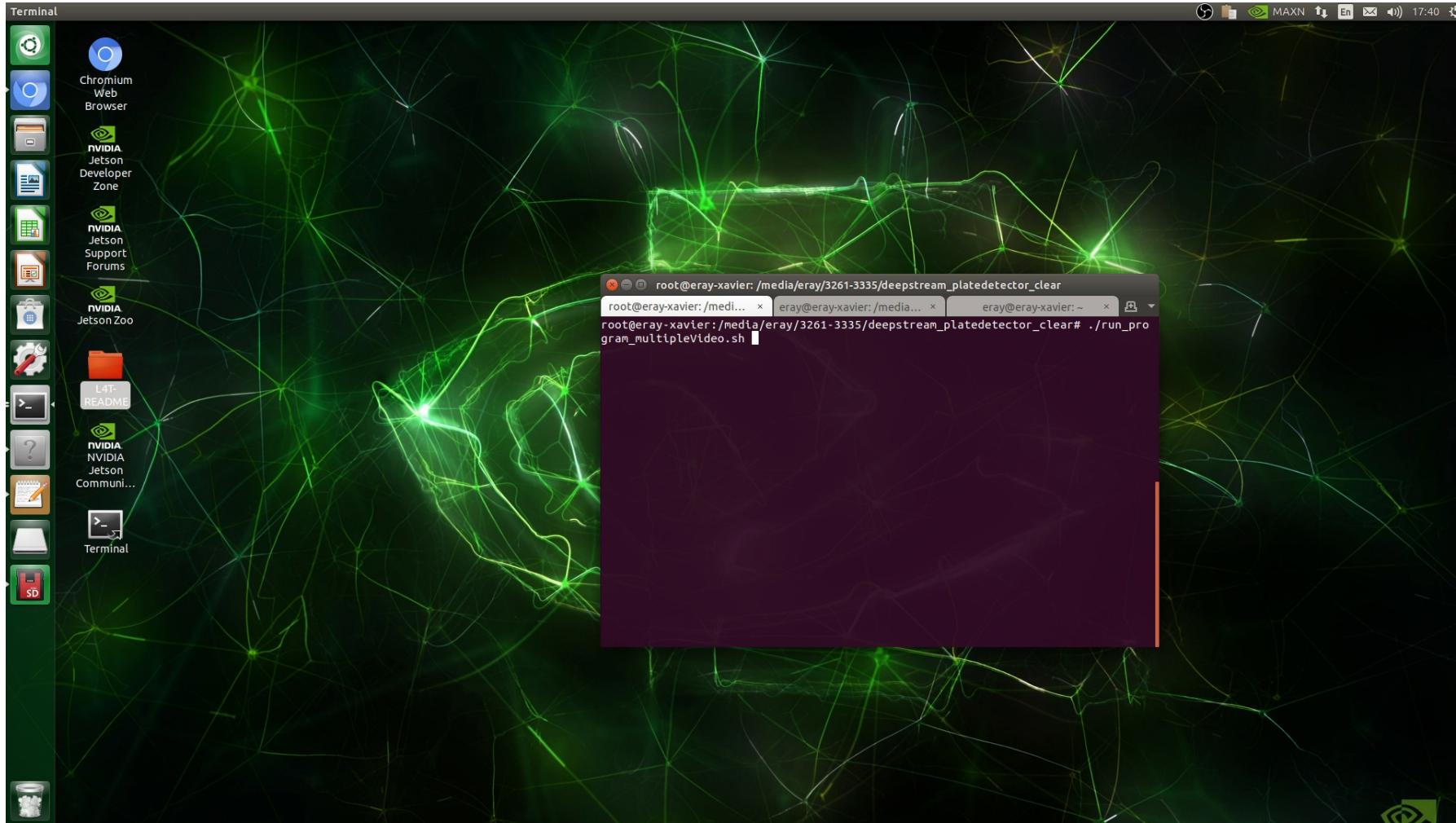


Xavier使用DeepStream(不含追蹤)





Xavier使用DeepStream(含追蹤)





Xavier使用DeepStream差異



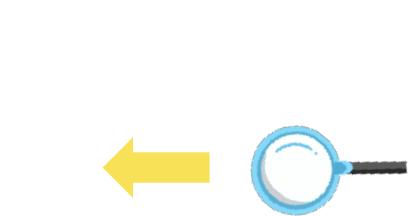
未使用 DeepStream	使用 DeepStream (不含追蹤)	使用 DeepStream (含追蹤)
16-20 FPS	54 FPS	60 FPS



Xavier使用TensorRT & DeepStream差異



未使用 DeepStream	使用TensorRT	使用 DeepStream (不含追蹤)	使用 DeepStream (含追蹤)
16-20 FPS	36-42 FPS	54 FPS	60 FPS



THANK YOU •••

