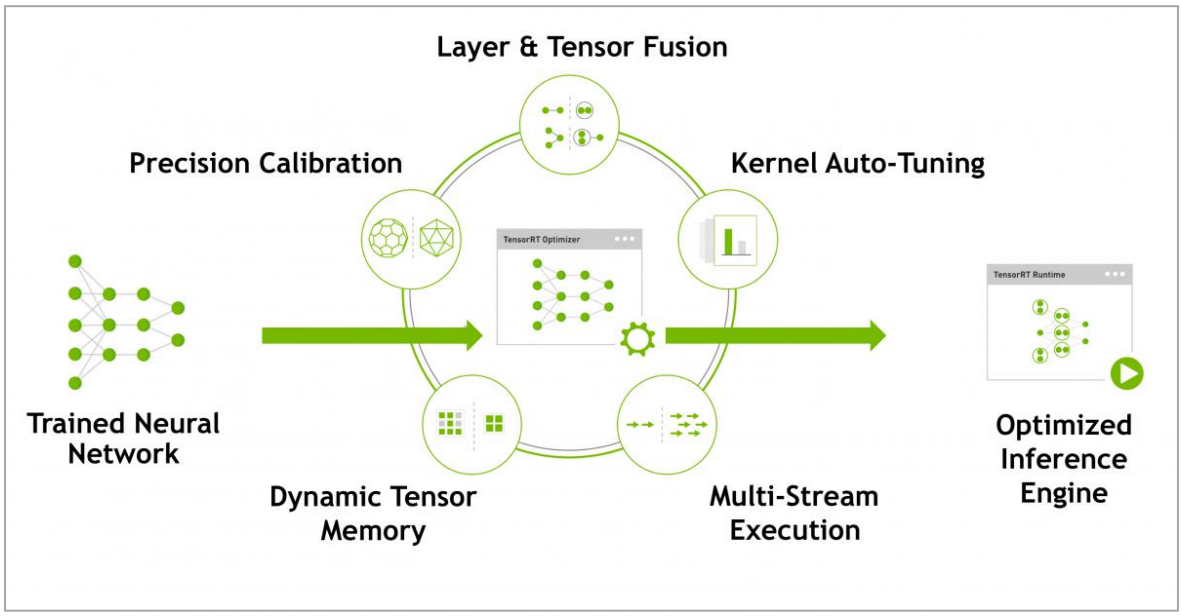
5.TensorRT Environmental construction （jetson-inference）



1. Preparation:

Because the model required for this example is about 1G, most of this example is not stored on the SD card (the SD card only has TensorRT required to run this model).

We need to remotely transfer the previously downloaded package to the corresponding directory.

First, if you don't have git and cmake, you need to install them.

```
sudo apt-get install libpython3-dev python3-numpy
sudo apt-get install git cmake
```

Clone the jetson-inference library

```
git clone https://github.com/dusty-nv/jetson-inference
```

Note:

If the system prompts, "error: RPC failed; curl 56 GnuTLS recv error (-54): Error in the pull function."

This reason is due to insufficient git default cache size.

We can increase the cache size using the following command

```
git config --global http.postBuffer 5242880000
```

If it still didn't work, it may be that the network speed is slow, configure the minimum speed and minimum speed time of git.

```
git config --global http.lowSpeedLimit 0
git config --global http.lowSpeedTime 999999
```

If it still does not work, it is a network problem in my own experience, and it is

recommended to do it when the network condition is better.

```
cd jetson-inference
git submodule update --init

mkdir build        # Create build folder
cd build           # Enter build
cmake ../          # Run cmake, it will automatically execute CMakePrebuild.sh in the upper directory
```

!!!Note:
During this period, the system may prompt whether you need to install pytorch, please choose according to your personal needs.

Proceed as follows:
Edit jetson-inference/CMakePrebuild.sh.
1)Comment out ./download-models.sh (with a # comment in front) as shown in Figure 1).
2) Transfer the jetson-inference package to the data/networks directory through WinSCP. As shown in Figure 2.

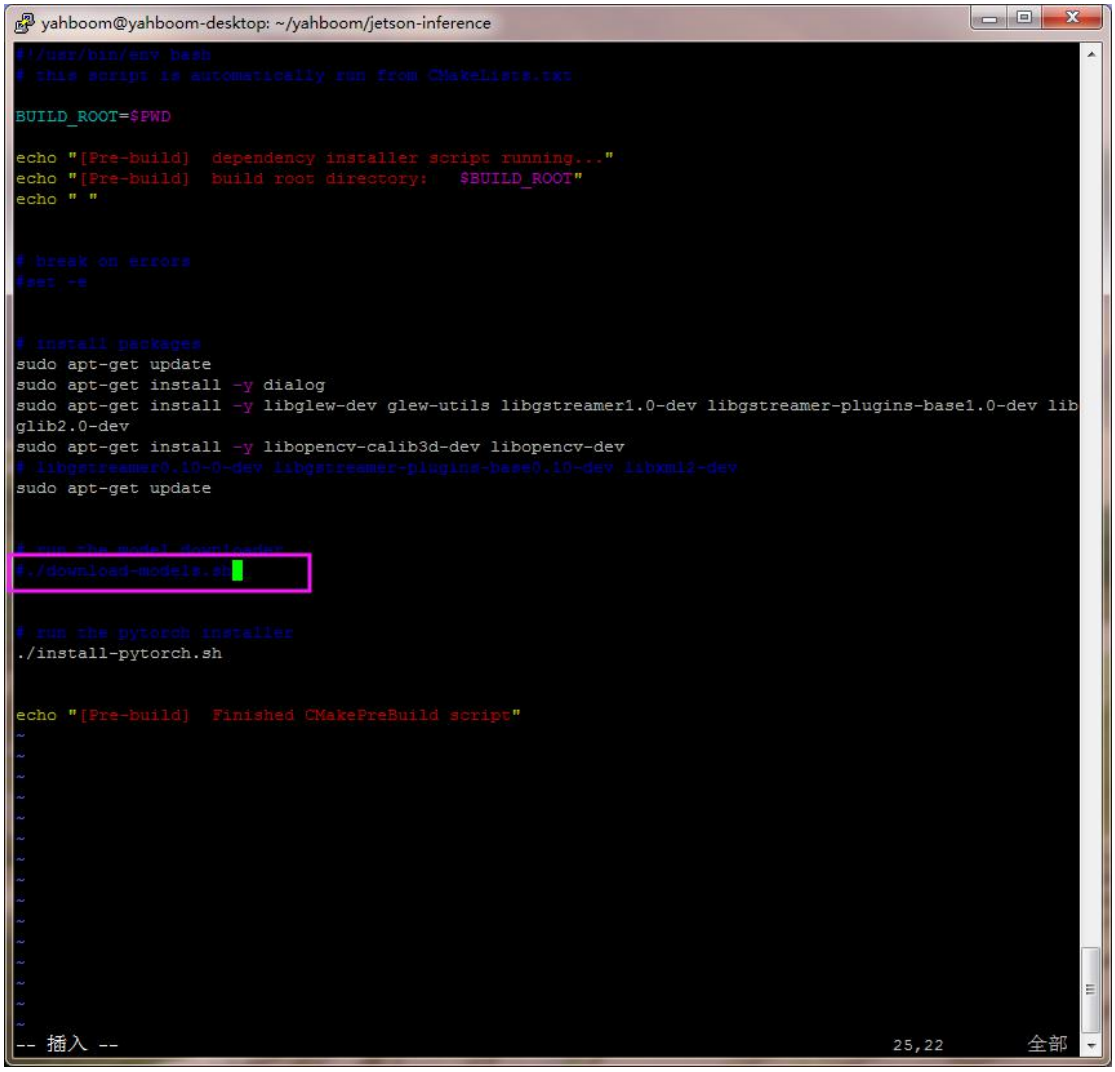


Figure 1

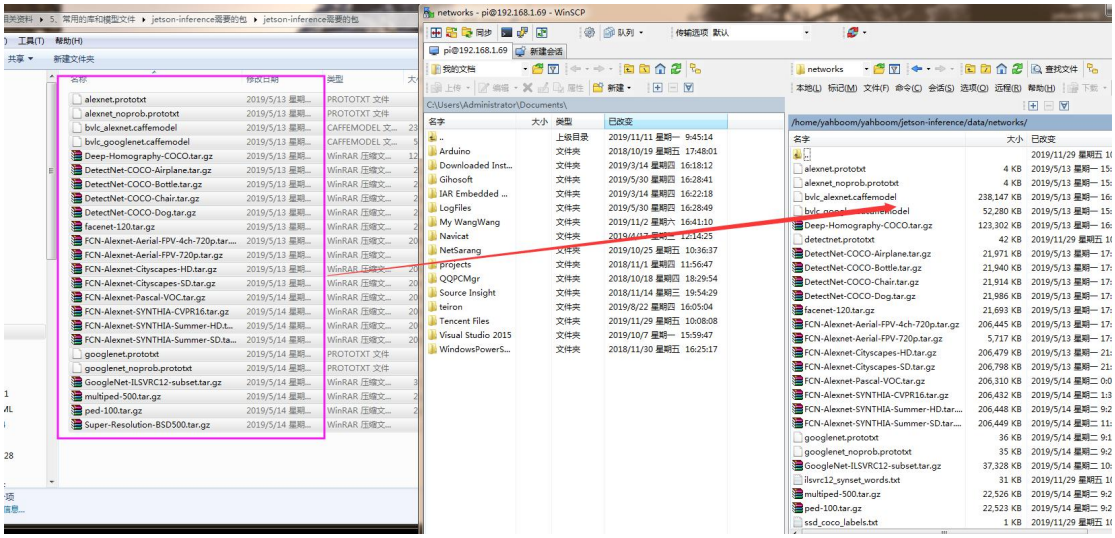


Figure 2

Input the following command to unzip.tar.gz files:

for tar in *.tar.gz; do tar xvf $tar; done

```
nano@nano-desktop:~/jetson-inference/data/networks$ ls
alexnet_noprob.prototxt                 FCN-Alexnet-Cityscapes-HD.tar.gz
alexnet.prototxt                        FCN-Alexnet-Cityscapes-SD
bvlc_alexnet.caffemodel                 FCN-Alexnet-Cityscapes-SD.tar.gz
bvlc_googlenet.caffemodel               FCN-Alexnet-Pascal-VOC
Deep-Homography-COCO                    FCN-Alexnet-Pascal-VOC.tar.gz
Deep-Homography-COCO.tar.gz             FCN-Alexnet-SYNTHIA-CVPR16
DetectNet-COCO-Airplane                 FCN-Alexnet-SYNTHIA-CVPR16.tar.gz
DetectNet-COCO-Airplane.tar.gz          FCN-Alexnet-SYNTHIA-Summer-HD
DetectNet-COCO-Bottle                   FCN-Alexnet-SYNTHIA-Summer-HD.tar.gz
DetectNet-COCO-Bottle.tar.gz            FCN-Alexnet-SYNTHIA-Summer-SD
DetectNet-COCO-Chair                    FCN-Alexnet-SYNTHIA-Summer-SD.tar.gz
DetectNet-COCO-Chair.tar.gz             GoogleNet-ILSVRC12-subset
DetectNet-COCO-Dog                      GoogleNet-ILSVRC12-subset.tar.gz
DetectNet-COCO-Dog.tar.gz               googlenet_noprob.prototxt
detectnet.prototxt                      googlenet.prototxt
facenet-120                             ilsvrc12_synset_words.txt
facenet-120.tar.gz                      multiped-500
FCN-Alexnet-Aerial-FPV-4ch-720p         multiped-500.tar.gz
FCN-Alexnet-Aerial-FPV-4ch-720p.tar.gz  ped-100
FCN-Alexnet-Aerial-FPV-720p             ped-100.tar.gz
FCN-Alexnet-Aerial-FPV-720p.tar.gz      Super-Resolution-BSD500
FCN-Alexnet-Cityscapes-HD               Super-Resolution-BSD500.tar.gz
nano@nano-desktop:~/jetson-inference/data/networks$
```

Input the following command to unzip.tar.gz files:

```
for gz in *.gz; do gunzip $gz; done
```

Input the following command to unzip multiple .tar.gz files:

```
for tar in *.tar.gz;    do tar xvf $tar; done
```

After cmake is successful, we need to compile

```
cd jetson-inference/build
# Make here without sudo
# Back -j4 uses 4 CPU cores to compile at the same time, reducing time
```

```
make (or make -j4)        Note: (In the build directory)
sudo make install         Note: (In the build directory)
```

If the compilation is successful, the following folder structure will be generated.

```
|-build
    \aarch64          (64-bit)
        \bin          where the sample binaries are built to
        \include      where the headers reside
        \lib          where the libraries are build to
    \armhf            (32-bit)
        \bin          where the sample binaries are built to
        \include      where the headers reside
        \lib          where the libraries are build to
```

3.Testing

We need to input the following command:

cd jetson-inference/build/aarch64/bin
./imagenet-console orange_0.jpg output_0.jpg

After waiting patiently, we will see the interface shown below.
(The first execution will take a long time)

```
yahboom@yahboom-desktop:~/yahboom/jetson-inference/build/aarch64/bin$ ./imagenet-console ./images/bird_
0.jpg output.jpg

imageNet -- loading classification network model from:
         -- prototxt     networks/googlenet.prototxt
         -- model        networks/bvlc_googlenet.caffemodel
         -- class_labels networks/ilsvrc12_synset_words.txt
         -- input_blob   'data'
         -- output_blob  'prob'
         -- batch_size   1
```

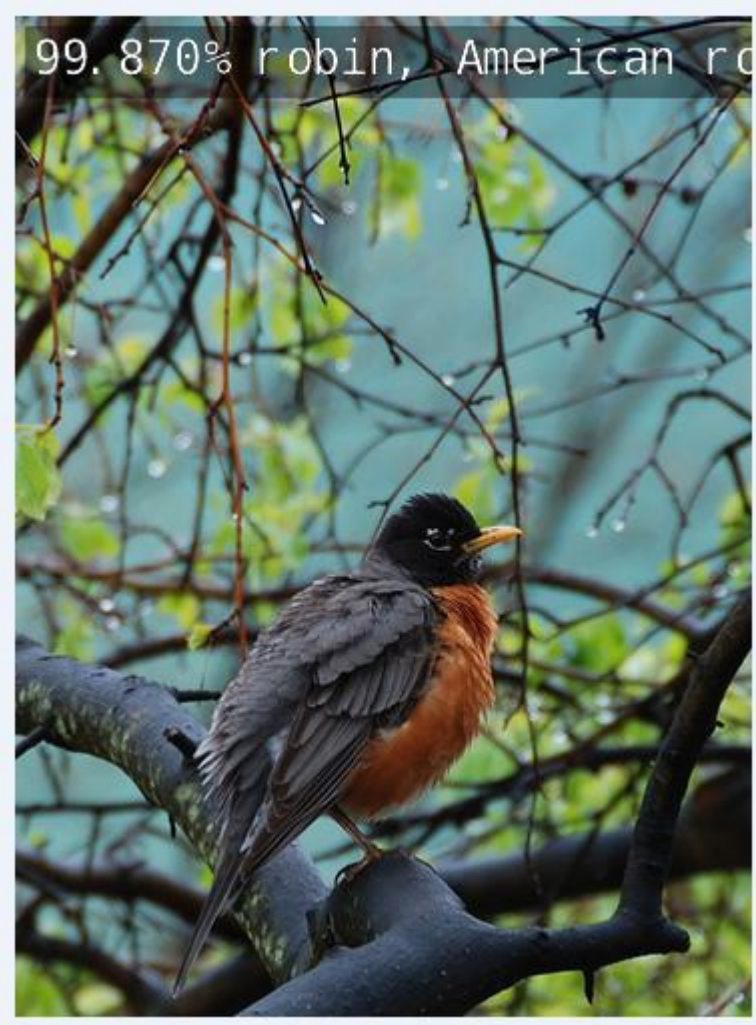
```
yahboom@yahboom-desktop: ~/yahboom/jetson-inference/build/aarch64/bin
                -- dim #0  3 (CHANNEL)
                -- dim #1  224 (SPATIAL)
                -- dim #2  224 (SPATIAL)
[TRT]   binding -- index   1
                -- name    'prob'
                -- type    FP32
                -- in/out  OUTPUT
                -- # dims  3
                -- dim #0  1000 (CHANNEL)
                -- dim #1  1 (SPATIAL)
                -- dim #2  1 (SPATIAL)
[TRT]   binding to input 0 data  binding index:  0
[TRT]   binding to input 0 data  dims (b=1 c=3 h=224 w=224) size=602112
[TRT]   binding to output 0 prob  binding index:  1
[TRT]   binding to output 0 prob  dims (b=1 c=1000 h=1 w=1) size=4000
device GPU, networks/bvlc_googlenet.caffemodel initialized.
[TRT]   networks/bvlc_googlenet.caffemodel loaded
imageNet -- loaded 1000 class info entries
networks/bvlc_googlenet.caffemodel initialized.
[image] loaded './images/bird_0.jpg'  (368 x 500, 3 channels)
class 0015 - 0.998702  (robin, American robin, Turdus migratorius)
imagenet-console:  './images/bird_0.jpg' -> 99.87018% class #15 (robin, American robin, Turdus migrator
ius)

[TRT]   ------------------------------------------------
[TRT]   Timing Report networks/bvlc_googlenet.caffemodel
[TRT]   ------------------------------------------------
[TRT]   Pre-Process   CPU   0.08995ms  CUDA   0.64693ms
[TRT]   Network       CPU  72.14478ms  CUDA  71.47083ms
[TRT]   Post-Process  CPU   0.97890ms  CUDA   1.06088ms
[TRT]   Total         CPU  73.21364ms  CUDA  73.17864ms
[TRT]   ------------------------------------------------

[TRT]   note -- when processing a single image, run 'sudo jetson_clocks' before
                to disable DVFS for more accurate profiling/timing measurements

imagenet-console:  attempting to save output image to 'output.jpg'
imagenet-console:  completed saving 'output.jpg'
imagenet-console:  shutting down...
imagenet-console:  shutdown complete
```

We need to find the corresponding directory to view **output_0.jpg**, the recognition result will be displayed at the top of the picture, as shown below.



For more detail, please see the official documentation:
Official Demo:
https://developer.nvidia.com/embedded/twodaystoademo
Official TensorRT tutorial:
https://docs.nvidia.com/deeplearning/sdk/tensorrt-sample-support-guide/index.html