Local deployment of yolo training environment

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Environment Configuration

In this section, we will demonstrate how to deploy the Yolo v5 environment locally and train a classification (recognition) model.

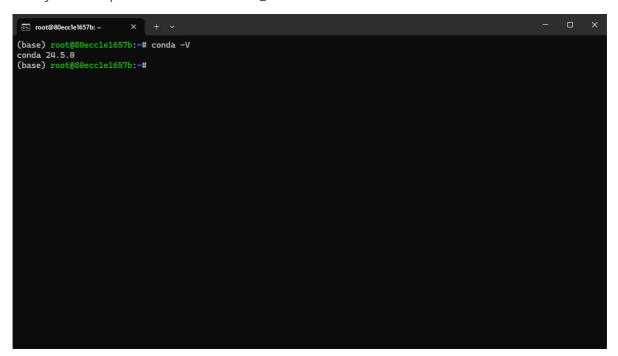
The operating environment we demonstrate is as follows:

System: Ubuntu 22.04

GPU: 4090D*1

CUDA version: 12.4

The system has pre-installed Miniconda3_3.10



Download and install Yolo v5 environment dependencies

git clone https://github.com/ultralytics/yolov5.git

```
(base) root@80eccle1657b:-# conda -V conda 24.5.0 (base) root@80eccle1657b:-# git clone https://github.com/ultralytics/yolov5.git 正克隆到 'yolov5'... remote: Enumerating objects: 17360, done. remote: Counting objects: 100% (52/52), done. remote: Counting objects: 100% (34/34), done. remote: Total 17360 (delta 36), reused 18 (delta 18), pack-reused 17308 (from 2) 接收对象中: 100% (17360/17360), 16.25 MB | 12.48 MB/s, 完成. 处理 delta 中: 100% (17901/11901), 完成. (base) root@80eccle1657b:-# |
```

implement

```
cd yolov5
pip install -r requirements.txt
```

This part of the file contains a lot of content, please be patient

```
Downloading ultralytics.thop-2.0.14-py3-none-any.whl (26 kB)
Downloading filelock-3.18.0-py3-none-any.whl (16 kB)
Downloading filelock-3.18.0-py3-none-any.whl (16 kB)
Downloading fisepec-2025.3.2-py3-none-any.whl (194 kB)

Downloading networkx-3.4.2-py3-none-any.whl (1.7 MB)

Downloading networkx-3.4.2-py3-none-any.whl (2.6 kB)
Downloading mpmath-1.3.0-py3-none-any.whl (22 kB)
Downloading smmap-5.0.2-py3-none-any.whl (536 kB)

Downloading smmap-5.0.2-py3-none-any.whl (24 kB)

Downloading smmap-5.0.2-py3-none-any.whl (24 kB)

Installing collected packages: triton, pytz, py-cpuinfo, nvidia-cusparselt-cul2, nvidia-cula-cul2, nvidia-cufft-cul2, nvidia-cuda-runtime-cul2, nvidia-cuda-runtime-cul2, nvidia-cuda-runtime-cul2, nvidia-cuda-runtime-cul2, nvidia-cuda-runtime-cul2, nvidia-cuda-runtime-cul2, midia-cuda-cul2, nvidia-cuda-cul2, gitdb, co ntourpy, nvidia-cusolver-cul2, matplotlib, gitpython, torch, seaborn, ultralytics-thop, torchvision, thop, ultralytics Attempting uninstall setuptools

Found existing installation: setuptools 69.5.1

Uninstalling setuptools-69.5.1:

Successfully installed contourpy-1.3.1 cycler-0.12.1 filelock-3.18.0 fonttools-4.57.0 fsspec-2025.3.2 gitdb-4.0.12 gitpy thon-3.1.4H kiwisolver-1.4.8 matplotlib-3.10.1 mpmath-1.3.0 networkx-3.4.2 numpy-2.1.1 nvidia-cudna-cul2-12.4.5.8 nvidia-cuda-cupti-cul2-12.4.127 nvidia-cuda-nvrtc-cul2-12.4.127 nvidia-cuda-cul2-12.4.127 nvidia-cudna-cul2-12.3.1.170 nvidia-cuda-cupti-cul2-12.4.127 nvidia-cuda-nvrtc-cul2-12.4.127 nvidia-cuda-cul2-12.2.4.127 nvidia-cuda-cul2-12.3.1.170 nvidia-cuda-cupti-cul2-12.4.127 nvidia-cuda-nvrtc-cul2-12.8.5 nvidia-nvijtilink-cul2-12.4.127 nvidia-cudan-cul2-12.3.1.170 nvidia-cuda-sester-cul2-11.5.2 seaborn-0.13.2 setupto ols-78.1.0 smmap-5.0.2 sympy-1.13.1 thop-0.1.1 post209072238 torch-2.6.0 torchvision-0.21.0 triton-3.2.0 tzdata-2025.2 ultralytics-8.3.103 ultralytics-thop-2.0.14

WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system package manager. It is recommende
```

Preparing training data

Here we use the training set officially provided by Canaan Technology as sample data

Please download the sample dataset provided. The sample dataset contains classification, detection and segmentation datasets for three types of fruits (apple, banana and orange). Unzip the dataset to yolov5 the directory and use it fruits_cls as the dataset for the fruit classification task.

We execute the following code

```
wget https://kendryte-download.canaan-
creative.com/developer/k230/yolo_files/datasets.zip
```

unzip datasets.zip

```
inflating: datasets/fruits_yolo/labels/val/orange_24.txt
inflating: datasets/fruits_yolo/labels/val/orange_27.txt
inflating: datasets/fruits_yolo/labels/val/orange_29.txt
inflating: datasets/fruits_yolo/labels/val/orange_3.txt
extracting: datasets/fruits_yolo/labels/val/orange_30.txt
inflating: datasets/fruits_yolo/labels/val/orange_31.txt
inflating: datasets/fruits_yolo/labels/val/orange_32.txt
inflating: datasets/fruits_yolo/labels/val/orange_35.txt
inflating: datasets/fruits_yolo/labels/val/orange_36.txt
extracting: datasets/fruits_yolo/labels/val/orange_36.txt
extracting: datasets/fruits_yolo/labels/val/orange_47.txt
extracting: datasets/fruits_yolo/labels/val/orange_99.txt
inflating: datasets/fruits_yolo/labels/val/orange_51.txt
extracting: datasets/fruits_yolo/labels/val/orange_51.txt
inflating: datasets/fruits_yolo/labels/val/orange_51.txt
extracting: datasets/fruits_yolo/labels/val/orange_51.txt
extracting: datasets/fruits_yolo/labels/val/orange_51.txt
extracting: datasets/fruits_yolo/labels/val/orange_61.txt
extracting: datasets/fruits_yolo/labels/val/orange_67.txt
extracting: datasets/fruits_yolo/labels/val/orange_73.txt
extracting: datasets/fruits
```

Use YOLOv5 to train a fruit classification model

yolov5 Execute the command in the directory and use yolov5 to train three types of fruit classification models:

We enter the yolov5 directory (~/yolov5 in the example)

Execute the following code to start training

```
python classify/train.py --model yolov5n-cls.pt --data datasets/fruits_cls --epochs 100 --batch-size 8 --imgsz 224 --device '0'
```

If an interface similar to the one below appears, it means that training has started.

After the training is successfully completed, the output is as follows

```
root@80eccle1657b: ~/yolov5 ×
                                                                                                                                                                                                          29/29 [00:01<00:00, 22.21it/s]
29/29 [00:01<00:00, 22.89it/s]
29/29 [00:01<00:00, 22.81it/s]
29/29 [00:00<00:00, 23.81it/s]
29/29 [00:00<00:00, 33.60it/s]
29/29 [00:00<00:00, 33.60it/s]
29/29 [00:01<00:00, 24.29it/s]
29/29 [00:01<00:00, 23.20it/s]
29/29 [00:01<00:00, 23.37it/s]
29/29 [00:01<00:00, 23.37it/s]
29/29 [00:01<00:00, 23.66it/s]
29/29 [00:01<00:00, 24.44it/s]
29/29 [00:01<00:00, 24.93it/s]
29/29 [00:01<00:00, 24.93it/s]
29/29 [00:01<00:00, 24.93it/s]
29/29 [00:01<00:00, 24.93it/s]
29/29 [00:01<00:00, 22.92it/s]
29/29 [00:01<00:00, 23.96it/s]
29/29 [00:01<00:00, 23.98it/s]
29/29 [00:01<00:00, 23.96it/s]
                                   0.21G
         81/100
                                                               0.292
                                                                                           0.329
                                                                                                                                                             1: 100%
                                                                                                                                                             1: 100%
1: 100%
         82/100
                                   0.21G
                                                               0.292
                                                                                           0.326
                                                                                                                                 1
                                  0.21G
                                                               0.292
                                                                                           0.332
         83/100
                                                                                                                                                             1: 100%
         84/100
                                  0.21G
                                                               0.292
                                                                                           0.323
         85/100
                                  0.21G
                                                                                           0.325
                                                                                                                                                             1: 100%
         86/100
                                0.214G
                                                               0.292
                                                                                           0.324
                                                                                                                                                             1: 100%
                                                              0.292
0.292
                                                                                                                                                             1: 100%
1: 100%
         87/100
                                0.216G
                                                                                           0.323
                                0.216G
                                                                                           0.327
          88/100
         89/100
                                0.216G
                                                               0.292
                                                                                           0.324
                                                                                                                                                             1: 100%
                                                                                                                                                             1: 100%
1: 100%
         90/100
                                0.2166
                                                               0.292
                                                                                           0.327
         91/100
                                0.216G
                                                               0.292
                                                                                           0.328
          92/100
                                0.216G
                                                                                           0.327
                                                                                                                                                             1: 100%
1: 100%
1: 100%
         93/100
                                0.216G
                                                               0.292
                                                                                           0.323
                                                              0.292
0.292
                                                                                           0.325
         94/100
                                0.216G
          95/100
                                 0.216G
                                                                                           0.325
         96/100
                                0.216G
                                                               0.292
                                                                                           0.329
                                                                                                                                                                    100%
                                                              0.292
0.292
                                                                                                                                                             1: 100%
         97/100
                                0.216G
                                                                                           0.323
         98/100
                                0.216G
                                                                                           0.324
                                                                                                                                                                    100%
         99/100
                                                                                                                                                                                                            29/29 [00:00<00:00, 32.84it/s]
       100/100
                                0.216G
                                                               0.292
                                                                                           0.324
                                                                                                                                                              1: 100%
Training complete (0.035 hours)
                                     python classify/predict.py --weights runs/train-cls/exp/weights/best.pt --source im.jpg
python classify/val.py --weights runs/train-cls/exp/weights/best.pt --data datasets/fruits_cls
python export.py --weights runs/train-cls/exp/weights/best.pt --include onnx
model = torch.hub.load('ultralytics/yolov5', 'custom', 'runs/train-cls/exp/weights/best.pt')
https://netron.app
Results saved to runs/train-cls/exp
Predict:
Validate:
PyTorch Hub:
Visualize:
(base) root@80ecc1e1657b:~/yolov5#
```

Convert the trained model to a kmodel that can be recognized by K230

The model we trained with yolov5 is best.pt with the suffix pt

In order to use it on K230, we need to convert it into a model in kmodel format

Download the conversion tool

Execute the following code to install the required tools

```
sudo apt-get install -y dotnet-sdk-7.0

pip install --upgrade pip

pip install nncase==2.9.0

pip install nncase-kpu==2.9.0

pip install onnx

pip install onnxruntime

pip install onnxsim
```

```
pypa.io/warnings/venv. Use the --root-user-action option if you know what you are doing and want to suppress this warning.

Collecting onnxsim
Downloading onnxsim=0.4.36-cp310-cp310-manylinux_2_17_x86_64.manylinux_2014_x86_64.whl.metadata (4.3 kB)
Requirement already satisfied: onnx in /root/miniconda3/lib/python3.10/site-packages (from onnxsim)
Downloading rich-14.0.0-py3-none-any.whl.metadata (18 kB)
Requirement already satisfied: numpy=2.1.20 in /root/miniconda3/lib/python3.10/site-packages (from onnx->onnxsim) (2.1.1)
Requirement already satisfied: pyposepsione-any.whl.metadata (18 kB)
Requirement already satisfied: pyosepsione-any.whl.metadata (6.9 kB)
Downloading markdown-it-py>=2.2.0 (from rich->onnxsim)
Downloading markdown-it-py>=3.0.0-py3-none-any.whl.metadata (6.9 kB)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in /root/miniconda3/lib/python3.10/site-packages (from rich->onnxsim)
Requirement already satisfied: typing-extensions<5.0,>=4.0.0 in /root/miniconda3/lib/python3.10/site-packages (from rich->onnxsim)
Collecting mdurl-=0.1 (from markdown-it-py>=2.2.0->rich->onnxsim)
Downloading mdurl-0.1.2-py3-none-any.whl.metadata (1.6 kB)
Downloading monxsim-0.4.36-cp310-cp310-manylinux_2-17_x86_64.manylinux2014_x86_64.whl (2.3 MB)
Downloading markdown.it-py-3.0.0-py3-none-any.whl (87 kB)
Downloading markdown.it-py-3.0.0-py3-none-any.whl (10.0 kB)
Installing collected packages: mdurl, markdown-it-py, rich, onnxsim
Successfully installed markdown-it-py-3.0.0 mdurl-0.1.2 onnxsim-0.4.36 rich-14.0.0
WARNING: Running pip as the 'root' user can result in broken permissions and conflicting behaviour with the system packa ge manager, possibly rendering your system unusable. It is recommended to use a virtual environment instead: https://pip.pypa.io/warnings/venv. Use the --root-user-action option if you know what you are doing and want to suppress this warning.
```

Download the script tool and test_yolov5.zip unzip the model conversion script tool to yolov5 the directory;

```
wget https://kendryte-download.canaan-
creative.com/developer/k230/yolo_files/test_yolov5.zip
unzip test_yolov5.zip
```

```
root@80ecc1e1657b: ~/yolov5 ×
  inflating: test_yolov5/test/banana_31.jpg
  inflating: test_yolov5/test/banana_32.jpg
inflating: test_yolov5/test/banana_39.jpg
  inflating: test_yolov5/test/banana_43.jpg
inflating: test_yolov5/test/banana_5.jpg
inflating: test_yolov5/test/banana_55.jpg
  inflating: test_yolov5/test/banana_65.jpg
  inflating: test_yolov5/test/banana_66.jpg
  inflating: test_yolov5/test/banana_74.jpg
   inflating: test_yolov5/test/banana_8.jpg
  inflating: test_yolov5/test/mixed_2.jpg
inflating: test_yolov5/test/mixed_22.jpg
inflating: test_yolov5/test/orange_10.jpg
  inflating: test_yolov5/test/orange_23.jpg
inflating: test_yolov5/test/orange_26.jpg
  inflating: test_yolov5/test/orange_38.jpg
   inflating: test_yolov5/test/orange_42.jpg
  inflating: test_yolov5/test/orange_52.jpg
inflating: test_yolov5/test/orange_59.jpg
inflating: test_yolov5/test/orange_6.jpg
  inflating: test_yolov5/test/orange_76.jpg
inflating: test_yolov5/test/orange_77.jpg
  inflating: test_yolov5/test/orange_82.jpg
  inflating: test_yolov5/test/orange_94.jpg
creating: test_yolov5/test_images/
inflating: test_yolov5/test_images/test.jpg
extracting: test_yolov5/test_images/test_apple.jpg
  inflating: test_yolov5/test_images/test_banana.jpg
  inflating: test_yolov5/test_images/test_orange.jpg
pase) root@80eccle1657b:~/yolov5#|
(base) root(
```

Start model format conversion

Step 1: Convert pt format to onnx format

```
python export.py --weight runs/train-cls/exp/weights/best.pt --imgsz 224 --batch
1 --include onnx
```

Note that "runs/train-cls/exp/weights/best.pt" here is the path of the pt model we trained.

If the previous steps are done exactly as the tutorial, then the path here does not need to be modified

Step 2: Enter the test yolov5/classify directory

```
cd test_yolov5/classify
```

```
root@80eccle1657b: ~/yolov5 × + ~
export: data=data/coco128.yaml, weights=['runs/train-cls/exp/weights/best.pt'], imgsz=[224], batch_size=1, device=cpu, h alf=False, inplace=False, keras=False, optimize=False, int8=False, per_tensor=False, dynamic=False, cache=, simplify=False, mlmodel=False, opset=17, verbose=False, workspace=4, nms=False, agnostic_nms=False, topk_per_class=100, topk_all=100, iou_thres=0.45, conf_thres=0.25, include=['onnx']
YOLOV5 / V7.0-411-gf4d8a84c Python-3.10.14 torch-2.6.0+cu124 CPU
Fusing layers...
Model summary: 117 layers, 1212307 parameters, 0 gradients, 2.9 GFLOPs
PyTorch: starting from runs/train-cls/exp/weights/best.pt with output shape (1, 3) (2.4 MB)
ONNX: starting export with onnx 1.17.0...
ONNX: export success ✓ 0.9s, saved as runs/train-cls/exp/weights/best.onnx (4.6 MB)
Export complete (1.4s)
python classify/predict.py --weights runs/train-cls/exp/weights/best.onnx

Validate: python classify/val.py --weights runs/train-cls/exp/weights/best.onnx

PyTorch Hub: model = torch.hub.load('ultralytics/yolov5', 'custom', 'runs/train-cls/exp/weights/best.onnx') # WARNI

NG A ClassificationModel not yet supported for PyTorch Hub AutoShape inference

Visualize: https://netron.app
Results saved to /root/yolov5/runs/train-cls/exp/weights
(base) root@80ecc1e1657b:~/yolov5# ls
                                                   export.py pyproject.toml
hubconf.py README.md
                                                                                                                                                                                  yolov5n-cls.pt
benchmarks.py
CITATION.cff
                                                                                                                                                   train.py
tutorial.ipynb
                                datasets.zip LICENSE
                                                                                README.zh-CN.md
CONTRIBUTING.md detect.py models requirements.txt
(base) root@80eccle1657b:~/yolov5# cd test_yolov5/
(base) root@80eccle1657b:~/yolov5/test_yolov5# cd classify/
                                                                                requirements.txt test_yolov5.zip val.py
(base) root@80ecc1e1657b:~/yolov5/test_yolov5/classify# |
```

```
python to_kmodel.py --target k230 --model ../../runs/train-
cls/exp/weights/best.onnx --dataset ../test --input_width 224 --input_height 224
--ptq_option 0
```

Note: An error may occur at this time: RuntimeError:

/root/miniconda3/bin/../lib/libstdc++.so.6: version `GLIBCXX_3.4.30' not found (required by /usr/lib/dotnet/host/fxr/7.0.19/libhostfxr.so): Success

Solution:

Check if GLIBCXX 3.4.30 exists

```
strings /usr/lib/x86_64-linux-gnu/libstdc++.so.6 | grep GLIBCXX
```

If there is one in the output, it means that the file is not really missing, but the version of libstdc++ in the virtual environment is lower than the version of the libstdc++ library in the system.

```
GLIBCXX_3.4.13
GLIBCXX_3.4.4
GLIBCXX_3.4.5
GLIBCXX_3.4.7
GLIBCXX_3.4.8
GLIBCXX_3.4.19
GLIBCXX_3.4.10
GLIBCXX_3.4.11
GLIBCXX_3.4.12
GLIBCXX_3.4.13
GLIBCXX_3.4.14
GLIBCXX_3.4.14
GLIBCXX_3.4.15
GLIBCXX_3.4.15
GLIBCXX_3.4.16
GLIBCXX_3.4.17
GLIBCXX_3.4.16
GLIBCXX_3.4.17
GLIBCXX_3.4.18
GLIBCXX_3.4.18
GLIBCXX_3.4.18
GLIBCXX_3.4.18
GLIBCXX_3.4.20
GLIBCXX_3.4.20
GLIBCXX_3.4.20
GLIBCXX_3.4.21
GLIBCXX_3.4.21
GLIBCXX_3.4.22
GLIBCXX_3.4.22
GLIBCXX_3.4.22
GLIBCXX_3.4.23
GLIBCXX_3.4.26
GLIBCXX_3.4.26
GLIBCXX_3.4.27
GLIBCXX_3.4.26
GLIBCXX_3.4.27
GLIBCXX_3.4.28
GLIBCXX_3.4.26
GLIBCXX_3.4.27
GLIBCXX_3.4.28
GLIBCXX_3.4.28
GLIBCXX_3.4.28
GLIBCXX_3.4.28
GLIBCXX_3.4.29
GLIBCX_3.4.29
```

At this point we only need to execute the following command

Delete the libstdc++.so.6 file in the virtual environment (or rename it to another name, here we rename it to libstdc++.so.6.old)

```
mv /root/miniconda3/bin/../lib/libstdc++.so.6
/root/miniconda3/bin/../lib/libstdc++.so.6.old
```

Then create a soft link and let the libstdc++.so.6 called by the virtual environment point to the libstdc++.so.6 that comes with the system.

```
ln -s /usr/lib/x86_64-linux-gnu/libstdc++.so.6 libstdc++.so.6
```

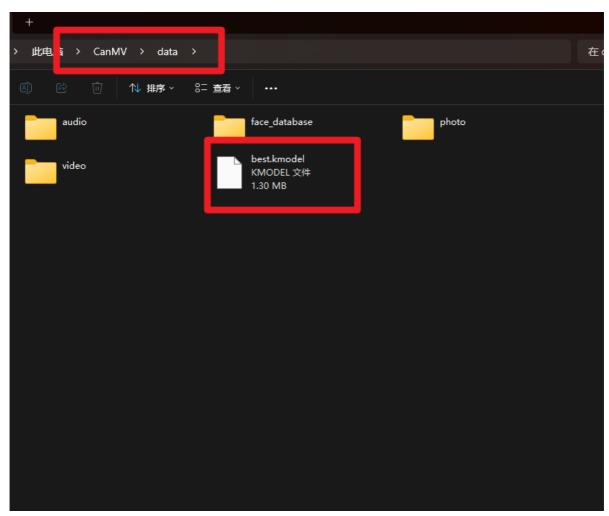
After the conversion command is completed, the output is as follows

The generated model is in the directory ~/yolov5/runs/train-cls/exp/weights

```
in/../lib/libstdc++.so.6.old
(base) root@80eccle1657b:-/yolov5/test_yolov5/classify# ln -s /usr/lib/x86_64-linux-gnu/libstdc++.so.6 libstdc++.so.6
(base) root@80eccle1657b:-/yolov5/test_yolov5/classify# python to_kmodel.py --target k230 --model ../../runs/train-cls/e xp/weights/best.onnx --dataset ../test --input_width 224 --input_height 224 --ptq.option 0
warn: Nncase.Hosting.Pluginloader[0]
NNCASE.PLUGIN_PATH is not set.
/root/yolov5/test_yolov5/classify/to_kmodel.py:25: DeprecationWarning: 'mapping_TENSOR_TYPE_TO_NP_TYPE' is now deprecate d and will be removed in a future release.To silence this warning, please use 'helper.tensor_dtype_to_np_dtype' instead. input_dict['dtype'] = onnx.mapping.TENSOR_TYPE_TO_NP_TYPE[onnx_type.elem_type]
WARNING: The argument 'input_shapes' is deprecated. Please use 'overwrite_input_shapes' and/or 'test_input_shapes' instead. An error will be raised in the future.
(base) root@80eccle1657b:-/yolov5/test_yolov5/classify# cd .././
(base) root@80eccle1657b:-/yolov5/test_yolov5/classify# cd ././/
(base) root@80eccle1657b:-/yolov5/test_yolov5/classify# cd .././
(base) root@80eccle1657b:-/yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_yolov5/test_y
```

The name is best.kmodel

We copy this best.kmodel to the data directory of our K230

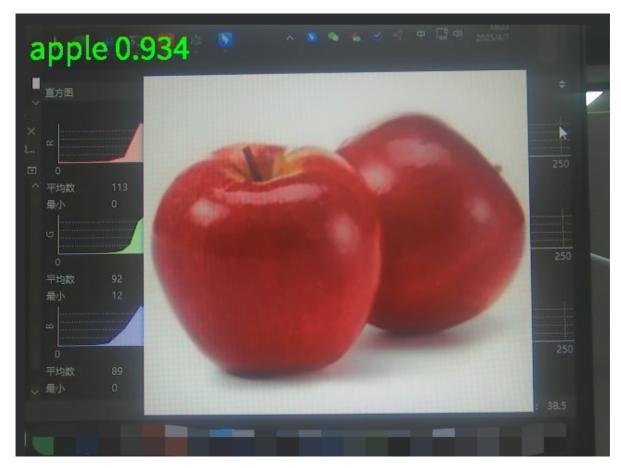


Then open CanMV IDE, copy the following code and click Run

```
from libs.YOLO import YOLOV5
import os,sys,gc
import ulab.numpy as np
import image
if __name__=="__main__":
   rgb888p_size=[1280,720]
   display_size=[640,480]
   # 模型路径 Model path
   kmodel_path="/data/best.kmodel"
   labels = ["apple","banana","orange"]
   confidence_threshold = 0.5
   model_input_size=[224,224]
   # 初始化PipeLine Initialize PipeLine
pl=PipeLine(rgb888p_size=rgb888p_size,display_size=display_size,display_mode="l
cd")
   pl.create()
   # 初始化YOLOv5实例 Initialize the YOLOv5 instance
yolo=YOLOv5(task_type="classify", mode="video", kmodel_path=kmodel_path, labels=la
bels,rgb888p_size=rgb888p_size,model_input_size=model_input_size,display_size=di
splay_size,conf_thresh=confidence_threshold,debug_mode=0)
   yolo.config_preprocess()
   try:
       while True:
           os.exitpoint()
           with ScopedTiming("total",1):
               # 逐帧推理 Frame-by-frame reasoning
               img = pl . get_frame ()
               res = yolo . run ( img )
               yolo . draw_result ( res , pl . osd_img )
               pl . show_image ()
               gc . collect ()
   except Exception as e:
       sys.print_exception ( e )
   finally:
       yolo.deinit ( )
       pl . destroy ()
```

Let's test the effect

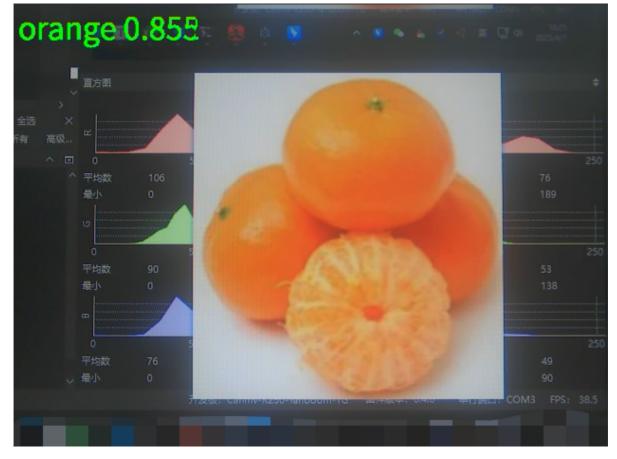
Identify Apples



Identify bananas



Identify oranges



You can see that this routine can run successfully.

Other considerations

- 1. Environment configuration, Yolov5 environment dependencies, and model conversion tools only need to be downloaded once, and no re-download is required for subsequent training
- 2. If the speed of pip download is much lower than the actual network bandwidth, please check whether the pip mirror source is configured.
- 3. The model trained with the current dataset may identify non-banana objects other than oranges/apples as bananas. This is normal. If a more accurate model is needed, a training set with a larger amount of data can be prepared for training.