

Human key point detection

Human key point detection

- Routine Experiment Effect

- Code Explanation

 - Code structure

 - Code Analysis

 - flow chart

- Brief description of human key point detection algorithm

 - Common application scenarios

 - Algorithm Overview

 - Network structure

Routine Experiment Effect

In this section, we will learn how to use K230 to realize the function of human key point detection.

The example code is in [Source code/08.Body/02.person_keypoint_detect.py]

After connecting to the IDE, run the sample code in this section, and use K230 to aim at a picture with multiple human bodies. You can see the positions of the key points of the human bodies marked on the screen. For scenes with multiple human bodies, they can also be identified more accurately.



Code Explanation

Code structure

Main program flow:

1. Initialization phase:
 - Initializing Pipeline
 - Initialize the PersonKeyPointApp class
 - Configuring preprocessing parameters
2. Main loop:
 - Get image frame
 - Perform model inference
 - Plotting the detection results
 - Displaying images
 - Performing garbage collection
3. Exception handling:
 - Detecting anomalies
 - If normal, continue the cycle
 - Exit the program if abnormal

Code Analysis

For the complete code, please refer to the file [Source Code/08.Body/02.person_keypoint_detect.py]

```
if __name__ == "__main__":
    # 显示模式, 默认"hdmi", 可以选择"hdmi"和"lcd"
    # Display mode, default "hdmi", can choose "hdmi" or "lcd"
    display_mode = "lcd"
    rgb888p_size = [640, 360] # RGB图像尺寸 (RGB image size)

    if display_mode == "hdmi":
        display_size = [640, 360] # HDMI显示尺寸 (HDMI display size)
    else:
        display_size = [640, 480] # LCD显示尺寸 (LCD display size)

    # 模型路径 (Model path)
    kmodel_path = "/sdcard/kmodel/yolov8n-pose.kmodel"
    # 其它参数设置 (Other parameter settings)
    confidence_threshold = 0.2 # 置信度阈值 (Confidence threshold)
    nms_threshold = 0.5 # 非极大值抑制阈值 (NMS threshold)

    # 初始化Pipeline (Initialize Pipeline)
    pl = Pipeline(rgb888p_size=rgb888p_size, display_size=display_size,
display_mode=display_mode)
    pl.create()

    # 初始化自定义人体关键点检测实例 (Initialize custom person keypoint detection
instance)
    person_kp = PersonKeyPointApp(
        kmodel_path,
        model_input_size=[320, 320],
        confidence_threshold=confidence_threshold,
        nms_threshold=nms_threshold,
        rgb888p_size=rgb888p_size,
        display_size=display_size,
```

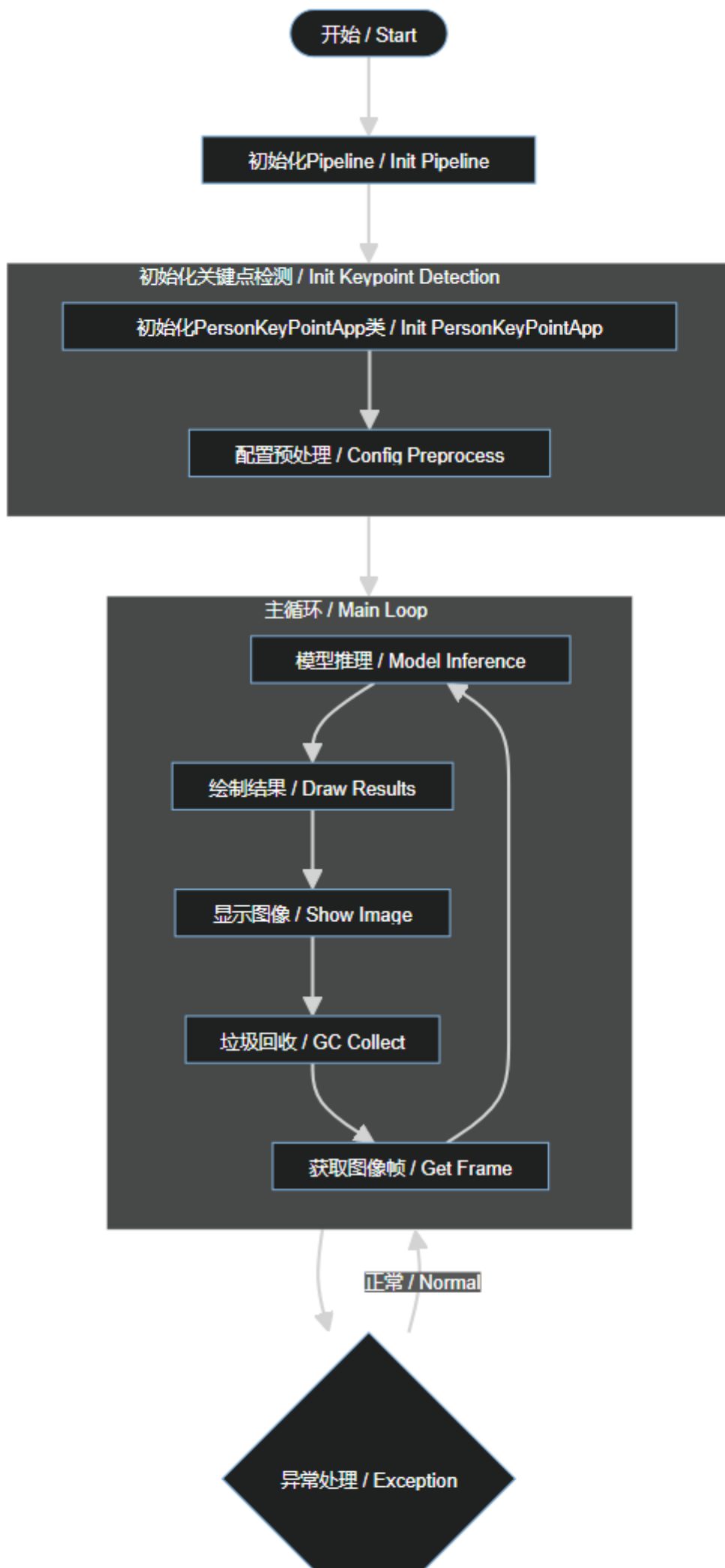
```
        debug_mode=0
    )

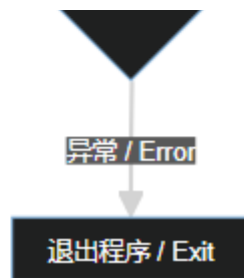
    person_kp.config_preprocess() # 配置预处理 (Configure preprocessing)

    while True:
        with ScopedTiming("total", 1):
            # 获取当前帧数据 (Get current frame data)
            img = pl.get_frame()
            # 推理当前帧 (Infer current frame)
            res = person_kp.run(img)
            # 绘制结果到PipeLine的osd图像 (Draw results to PipeLine's osd image)
            person_kp.draw_result(pl, res)
            # 显示当前的绘制结果 (Show current drawing results)
            pl.show_image()
            gc.collect() # 垃圾回收 (Garbage collection)

        # 释放资源 (Release resources)
        person_kp.deinit()
        pl.destroy()
```

flow chart





Brief description of human key point detection algorithm

Common application scenarios

Smart Fitness and Sports Analysis

- Posture correction and movement guidance
- Sports performance analysis and evaluation
- Remote Fitness Coaching System

Medical rehabilitation

- Patient motor function assessment
- Rehabilitation training progress tracking
- Telemedicine guidance

Security monitoring

- Abnormal behavior identification
- Fall detection and warning
- Crowd density and flow analysis

Human-computer interaction

- Gesture recognition control
- Virtual reality/augmented reality interaction
- Motion capture and 3D modeling

Sports

- Athlete technical movement analysis
- Game tactics research
- Training assistance system

Industrial production

- Work environment safety monitoring
- Workers' operational compliance inspection
- Human-machine collaborative control

Algorithm Overview

Network structure

Human key point detection usually adopts the following network architecture:

Backbone

- Use deep convolutional neural networks (such as ResNet, HRNet, etc.) to extract image features

- Capture human feature information at different scales through multi-scale feature extraction

Neck

- Feature pyramid structure (FPN) integrates features at different levels
- Combining upsampling and downsampling to enhance feature expression capabilities

Prediction Head

- Generate a heatmap to predict the location probability of each key point
- The number of output channels is equal to the number of key points (such as 17 key points in the COCO dataset)