# Self-learning object recognition

#### **Self-learning object recognition**

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# **Routine Experiment Effect**

In this section, we will learn how to use K230 to simply save the feature data of an object and perform basic recognition.

Let's take "mobile phone" and "ultrasonic module" as examples.

Run the example program in this section, place the [mobile phone] and [ultrasonic module] in front of the K230 camera respectively, and the K230 will start to record the features of the object at the center of the camera and save it as a binary bin file.

The current routine has added serial port output

For the protocol format, please refer to [Routine Communication Protocol.xlsx] in the document.

# **Code Explanation**

## Code structure

Initialization phase:

- Loading tracking templates and live models
- Initialize the tracker
- Set preprocessing parameters
- Initialize the AI2D preprocessor
- Configuring Tracking Thresholds

### Main loop phase:

- Image acquisition and preprocessing
- Template reasoning and real-time reasoning
- Tracking algorithm processing
- Results
- Resource Recycling

### Exit Processing:

- Abnormality Check
- Resource Cleanup

## **Key Code**

For the complete code, please refer to the file [Source Code/09.Scene/05.self\_learning.py]

```
# 自学习例程执行函数
# Self-learning demo execution function
def exce_demo(pl, recong_only=False):
   自学习例程主要流程:
     1. 初始化自学习实例
     2. 启动图像采集、推理、特征采集及后续匹配展示
   Main execution flow for self-learning demo:
     1. Initialize self-learning instance.
     2. Start continuous frame capture, inference, feature collection, and
matching display.
   参数:
     pl (PipeLine): PipeLine实例,用于图像采集和显示 // Pipeline instance for frame
capture and display.
     recong_only (bool): 是否仅作为识别模式,不进行特征采集 // Whether to run in
recognition-only mode.
   0.00
   global sl, database_path
   # 获取显示和图像参数
   # Retrieve display and image sizes
   display_mode = pl.display_mode
   rgb888p_size = pl.rgb888p_size
   display_size = pl.display_size
   # 模型路径设置
   # Set model path
   kmodel_path = "/sdcard/kmodel/recognition.kmodel"
   # 特征存储路径
   # Database (feature storage) path.
   database_path = "/sdcard/utils/features/"
   # 其它参数设置,例如模型输入尺寸、类别标签、top_k以及匹配阈值
   # Other parameters: model input size, labels, top_k and threshold.
   model_input_size = [224,224]
   labels = ["耳机", "超声波模块"]
   # labels = ["earphone", "ultrasonic"]
   top_k = 3
   threshold = 0.45
   try:
       # 初始化自学习实例
       # Initialize the SelfLearningApp instance
       s1 = SelfLearningApp(
           kmodel_path,
           model_input_size=model_input_size,
           labels=labels,
           top_k=top_k,
           threshold=threshold,
           database_path=database_path,
           rgb888p_size=rgb888p_size,
           display_size=display_size,
           debug_mode=0,
```

```
recong_only=recong_only
       )
       # 配置预处理操作
       # Configure the preprocessing operations
       sl.config_preprocess()
       # 无限循环处理每一帧数据
       # Infinite loop to process each frame
       while True:
           with ScopedTiming("total", 1):
               # 获取当前帧图像数据
               # Capture the current frame from the sensor
               img = pl.get_frame()
               # 推理当前帧图像,得到输出特征
               # Run inference on the current frame to get features
               res = sl.run(img)
               # 绘制结果到Pipeline的OSD图像中(包含特征采集框等信息)
               # Draw the results (feature collection box and matching info)
onto the pipeline's OSD image.
               sl.draw_result(pl, res)
               # 显示当前的图像及绘制的OSD信息
               # Display the image with the drawn OSD information.
               pl.show_image()
               # 主动调用垃圾回收,释放内存
               # Explicitly call garbage collection to free memory.
               gc.collect()
   except Exception as e:
       # 捕获异常后打印提示信息
       # Print message when an exception is caught.
       print("自学习例程退出")
   finally:
       # 最终退出时执行反初始化
       # Finally, deinitialize the demo.
       exit_demo()
# 退出例程,反初始化资源并清除特征保存文件夹
# Exit routine: cleanup resources and remove stored feature files
def exit_demo():
   global sl, database_path
   # 删除features文件夹下所有文件,并删除文件夹本身
   # Remove all files within the features directory and the directory itself
   stat_info = os.stat(database_path)
   # 判断目录是否为文件夹(目录标志位0x4000)
   # Check if database_path is a directory based on stat info (0x4000 indicates
a directory)
   if (stat_info[0] & 0x4000):
       list_files = os.listdir(database_path)
       for 1 in list_files:
           os.remove(database_path + 1)
   os.rmdir(database_path)
   # 执行自学习实例的反初始化,释放资源
   # Deinitialize the self-learning instance and free up resources.
   sl.deinit()
```

flow chart





