步骤 1 Tensoflow 搭建 Lenet 神经网络

1 搭建网络

```
label  8  has  2931  data
label  9  has  2907  data
```

```
In [3]: model = tf.keras.models.Sequential([
            tf.keras.layers.Conv2D(6, (5,5), activation='relu', input_shape=(32, 32, 1)),
            tf.keras.layers.MaxPooling2D(2, 2),

            tf.keras.layers.Conv2D(16, (5,5), activation='relu'),
            tf.keras.layers.MaxPooling2D(2,2),

            tf.keras.layers.Flatten(),
            tf.keras.layers.Dense(84, activation='relu'),
            tf.keras.layers.Dense(65, activation='softmax')
        ])
        model.summary()
```

2 载入数据集

```
In [3]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
        train_datagen = ImageDataGenerator(rescale=1./255)
        validation_datagen = ImageDataGenerator(rescale=1./255)

        train_generator = train_datagen.flow_from_directory(
                'train',
                color_mode='grayscale',
                target_size=(32, 32),
                batch_size=128,
                class_mode='categorical')
        validation_generator = validation_datagen.flow_from_directory(
                'test',
                color_mode='grayscale',
                target_size=(32, 32),
                batch_size=32,
                class_mode='categorical')
```

```
Found 52063 images belonging to 65 classes.
Found 5520 images belonging to 65 classes.
```

3 开始训练

```
In [9]: history = model.fit(
                train_generator,
                steps_per_epoch=8,
                epochs=100,
                verbose=1,
                validation_data = validation_generator,
                validation_steps=8)
```

```
8/8 [==============================] - 11s 1s/step - loss: 0.2013 - accuracy: 0.9512 - val_loss: 0.2409 - val_accuracy: 0.9414
Epoch 53/100
8/8 [==============================] - 8s 1s/step - loss: 0.2109 - accuracy: 0.9453 - val_loss: 0.2272 - val_accuracy: 0.9258
Epoch 54/100
8/8 [==============================] - 13s 2s/step - loss: 0.2104 - accuracy: 0.9512 - val_loss: 0.2676 - val_accuracy: 0.9414
Epoch 55/100
8/8 [==============================] - 7s 815ms/step - loss: 0.1630 - accuracy: 0.9502 - val_loss: 0.1698 - val_accuracy: 0.9688
Epoch 56/100
8/8 [==============================] - 8s 974ms/step - loss: 0.2052 - accuracy: 0.9561 - val_loss: 0.1672 - val_accuracy: 0.9531
Epoch 57/100
```

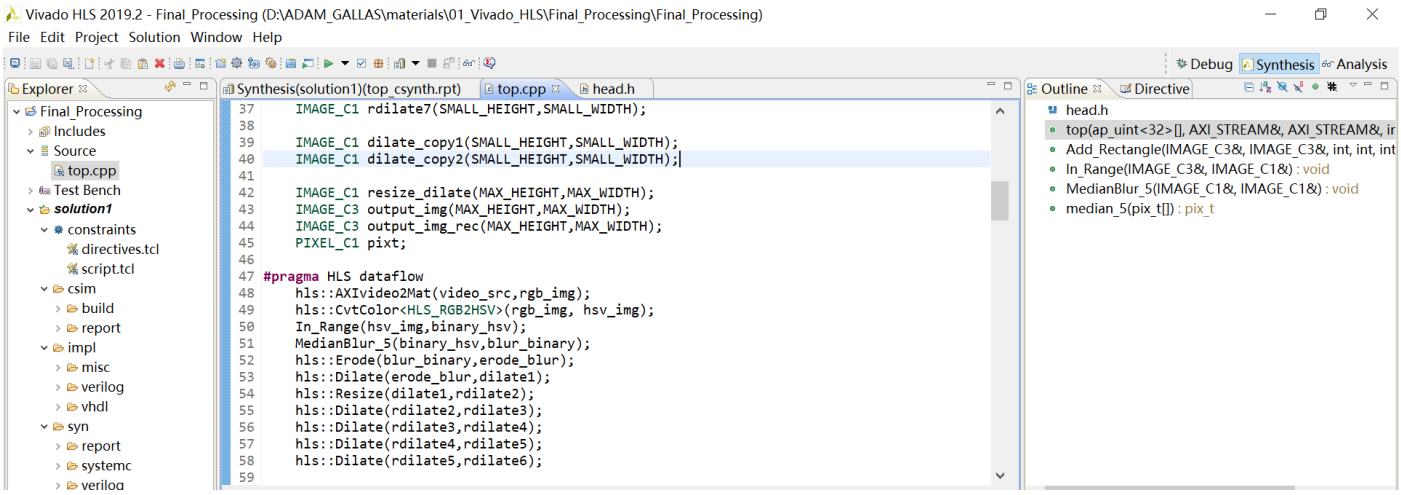
4 训练完成，导出模型

```
In [4]: bias1.tofile("hls_bias1.txt",sep=',\n',format='%s')
        bias2.tofile("hls_bias2.txt",sep=',\n',format='%s')
        bias3.tofile("hls_bias3.txt",sep=',\n',format='%s')
        bias4.tofile("hls_bias4.txt",sep=',\n',format='%s')
        weights1.tofile("hls_weights1.txt",sep=',\n',format='%s')
        weights2.tofile("hls_weights2.txt",sep=',\n',format='%s')
```

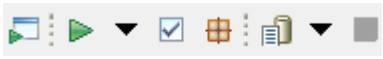
```
In [6]: hls_filter1=np.swapaxes(filter1,0,2)
        hls_filter1=np.swapaxes(hls_filter1,1,3)
        hls_filter1=np.swapaxes(hls_filter1,0,1)
```

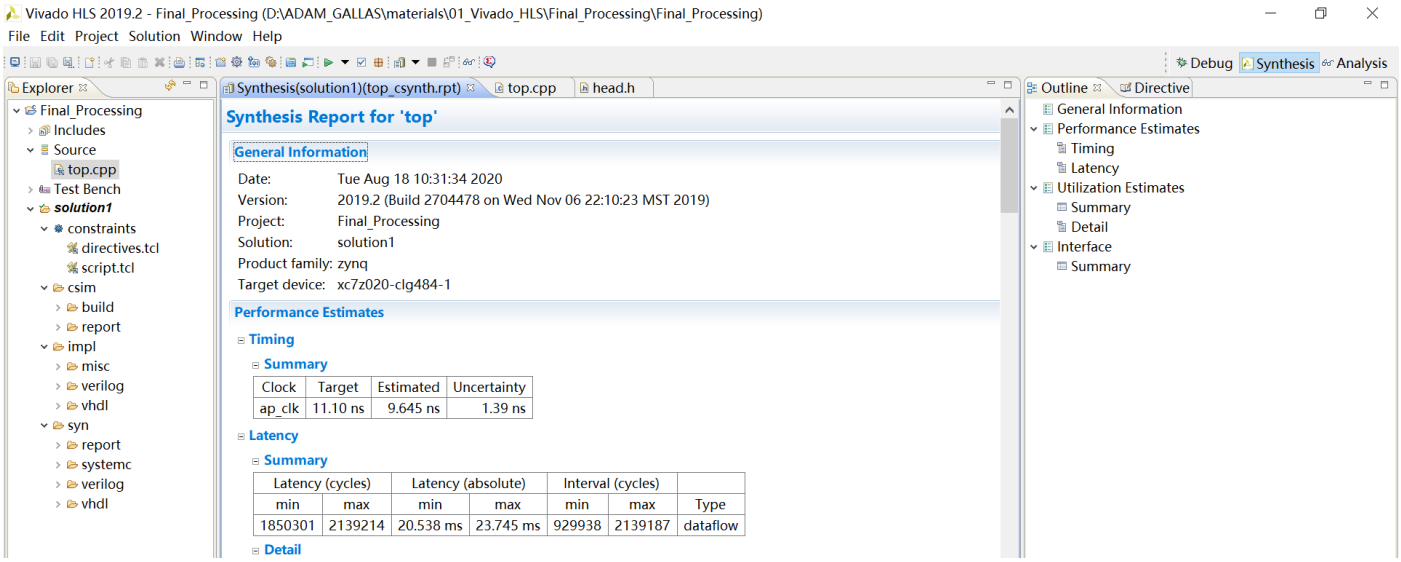
步骤 2 Vivado HLS 设计图像处理算法

1 设计算法

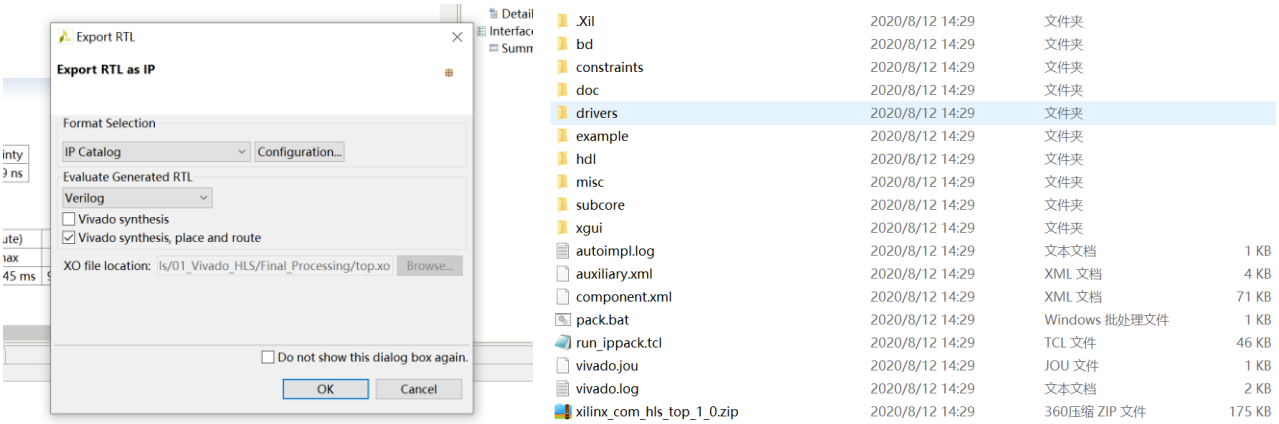


2 进行 C Simulation，Synthesis，C/RTL Cosimulation



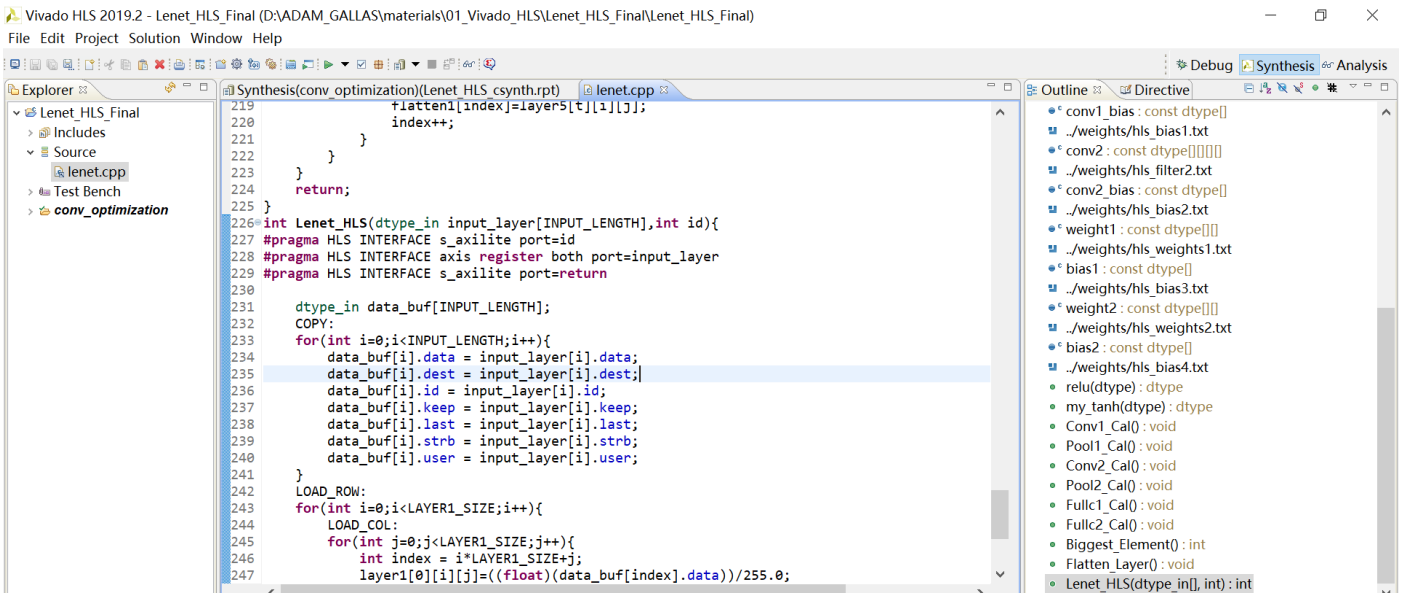


3 导出 IP 核



步骤 3 Vivado HLS 设计神经网络推理计算程序

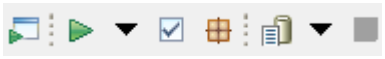
1 编写程序

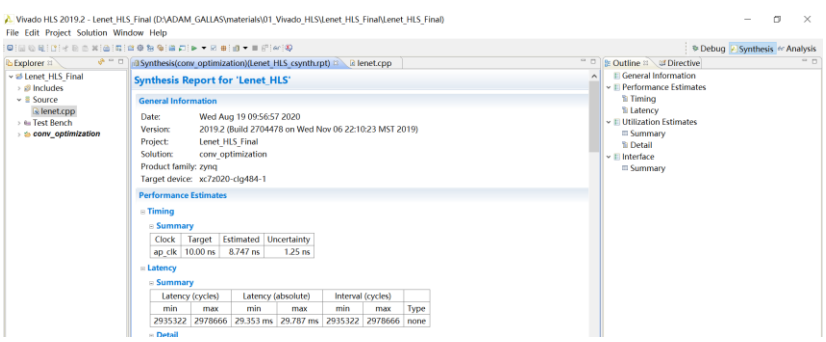


2 添加并行优化

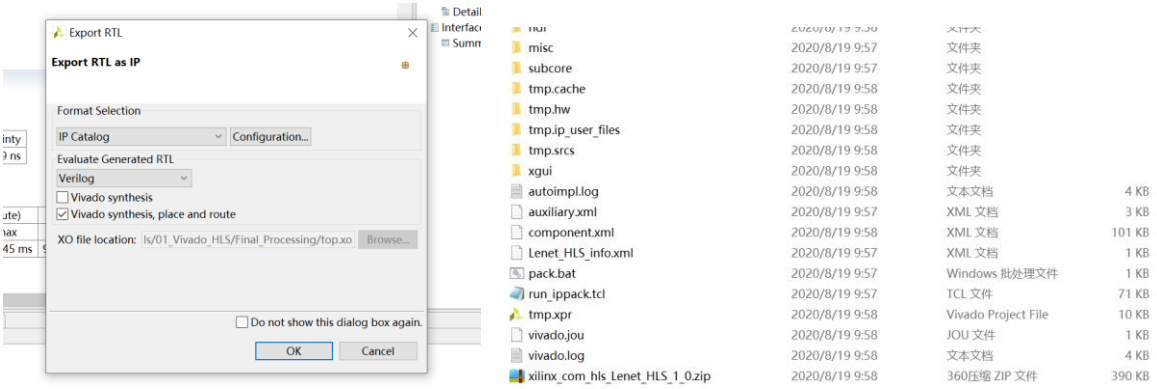
```
void Conv1_Cal(){

#pragma HLS ARRAY_PARTITION variable=conv1 complete dim=2
#pragma HLS ARRAY_PARTITION variable=conv1 complete dim=1
#pragma HLS ARRAY_PARTITION variable=layer2 complete dim=1
#pragma HLS ARRAY_PARTITION variable=layer1 complete dim=1

    CONV1_SIZE1:
    for(int i=0;i<CONV1_SIZE;i++){
        CONV1_SIZE2:
        for(int j=0;j<CONV1_SIZE;j++){
            CONV1_ROW:
            for(int row=0;row<LAYER2_SIZE;row++){
                CONV1_COL:
                for(int col=0;col<LAYER2_SIZE;col++){
#pragma HLS PIPELINE
                    CONV1_OUTD:
                    for(int out_d=0;out_d<LAYER2_DEPTH;out_d++){
                        if(i==0&&j==0){
                            layer2[out_d][row][col]=layer1[0][row+i][col+j]*conv1[out_d][0
                        }
                        else{
                            layer2[out_d][row][col]+=layer1[0][row+i][col+j]*conv1[out_d][
                        }
```

3 进行 C Simulation，Synthesis，C/RTL Cosimulation





4 导出 IP 核



步骤 4 Verilog 编写摄像头驱动程序

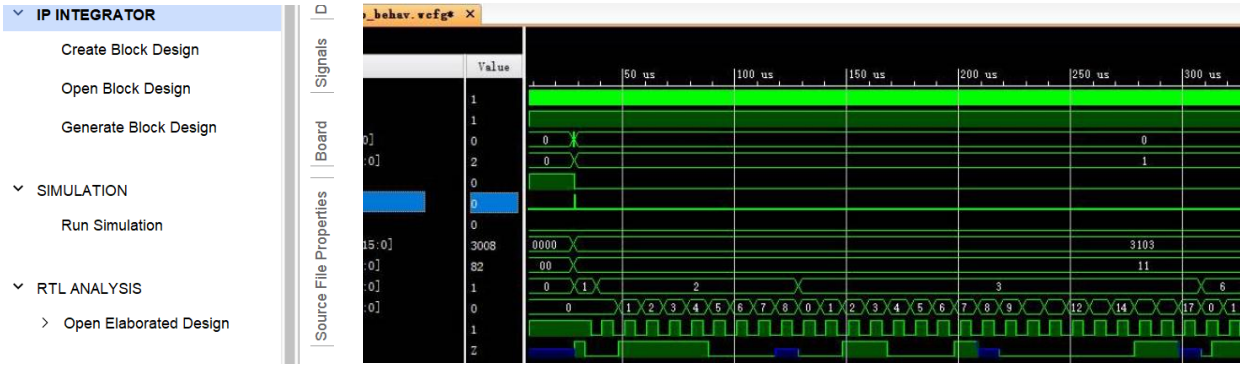
1 编写程序

Diagram | Address Editor | OV5640_DRIVE.v

D:/Project/ov_carplate/Vivado/rtl_ov5640_driver/OV5640_DRIVE.v

```
1  `timescale 1ns / 1ps
2  module OV5640_DRIVE(
3      input cmos_clk_i,//cmos senseor clock.
4      input rst_n_i,//system reset.active low.
5      input cmos_pclk_i,//input pixel clock.
6      input cmos_href_i,//input pixel hs signal.
7      input cmos_vsync_i,//input pixel vs signal.
8      input [7:0]cmos_data_i,//data.
9      output                cam_rst_n   ,
10     output                cam_pwdn    ,
11     output cmos_xclk_o,//output clock to cmos sensor.
12     output [23:0] rgb_o,
13     output clk_ce,
14     output de_o,
15     output vs_o,
16     output hs_o
17     );
18
19 assign cmos_xclk_o = cmos_clk_i;
20 parameter[3:0]CMOS_FRAME_WAITCNT = 4'd15;
21 reg[4:0] rst_n_reg = 5'd0;
22 reg cmos_href_r;
```

2 时序仿真

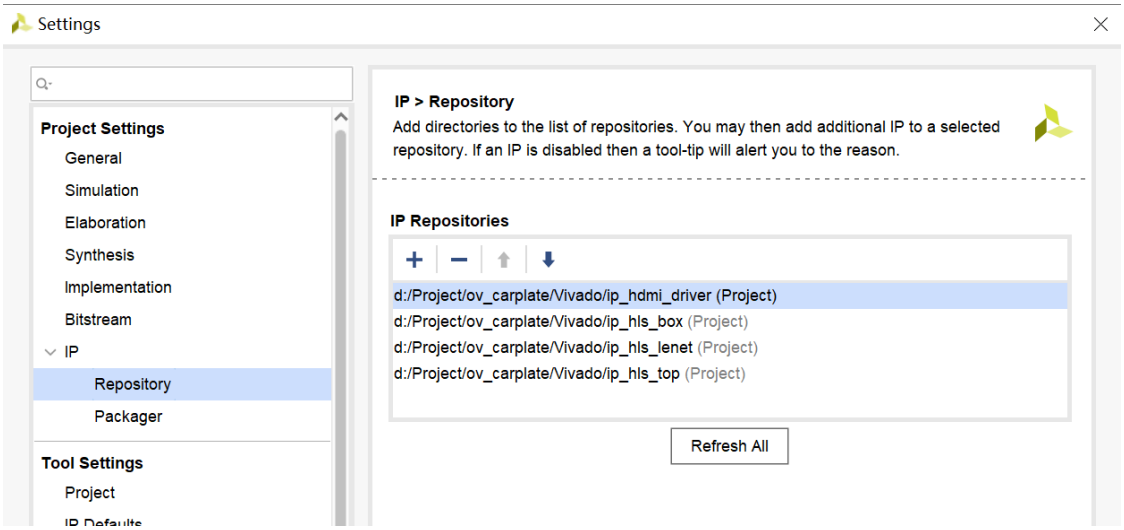


步骤 5 Vivado Block Design 设计

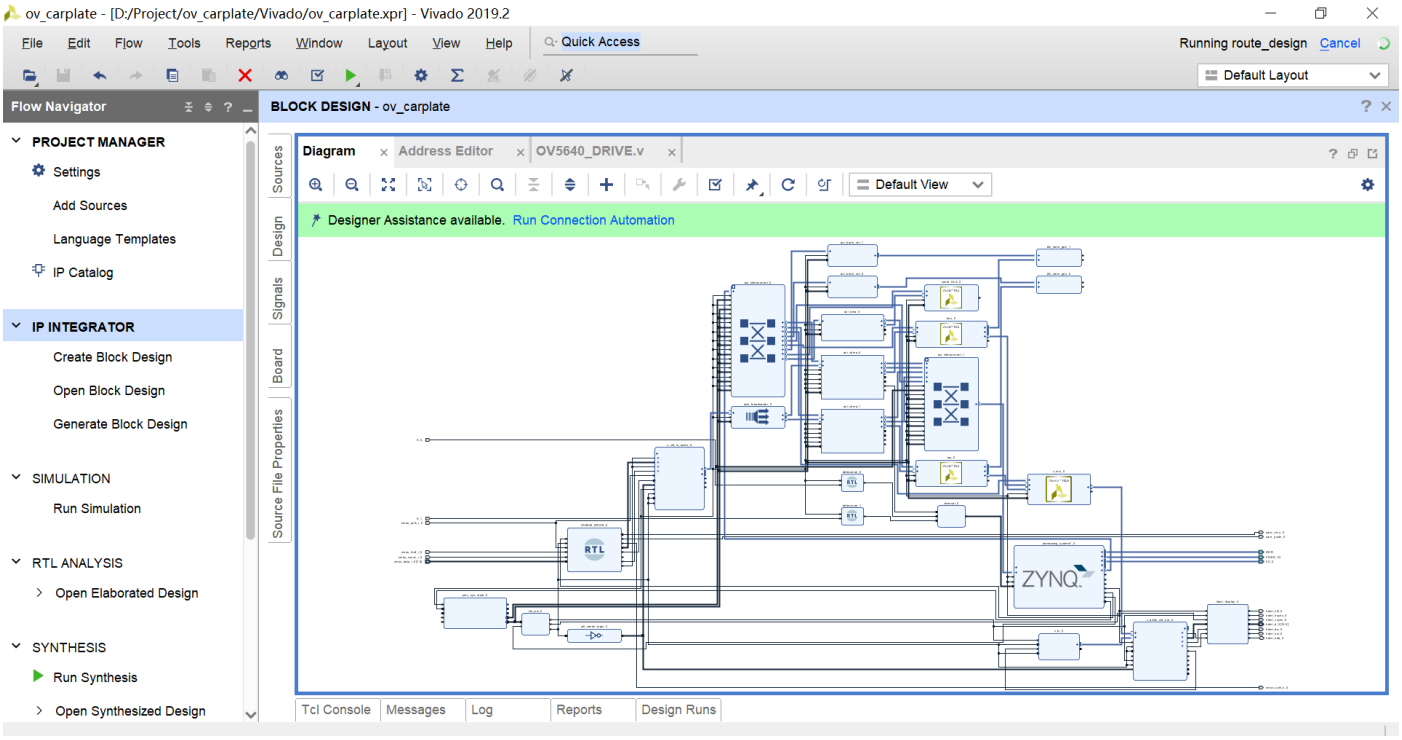
1 汇集 RTL 模块与 IP 核

ip_hdmi_driver	2020/8/24 10:06	文件夹
ip_hls_box	2020/8/24 11:11	文件夹
ip_hls_lenet	2020/8/24 11:11	文件夹
ip_hls_top	2020/8/24 11:11	文件夹
ov_carplate.cache	2020/8/24 13:54	文件夹
ov_carplate.hw	2020/8/24 11:07	文件夹
ov_carplate.ip_user_files	2020/8/24 13:55	文件夹
ov_carplate.runs	2020/8/24 13:58	文件夹
ov_carplate.sim	2020/8/24 11:07	文件夹
ov_carplate.srcs	2020/8/24 13:46	文件夹
rtl_button_driver	2020/8/24 10:06	文件夹
rtl_ov5640_driver	2020/8/24 10:06	文件夹

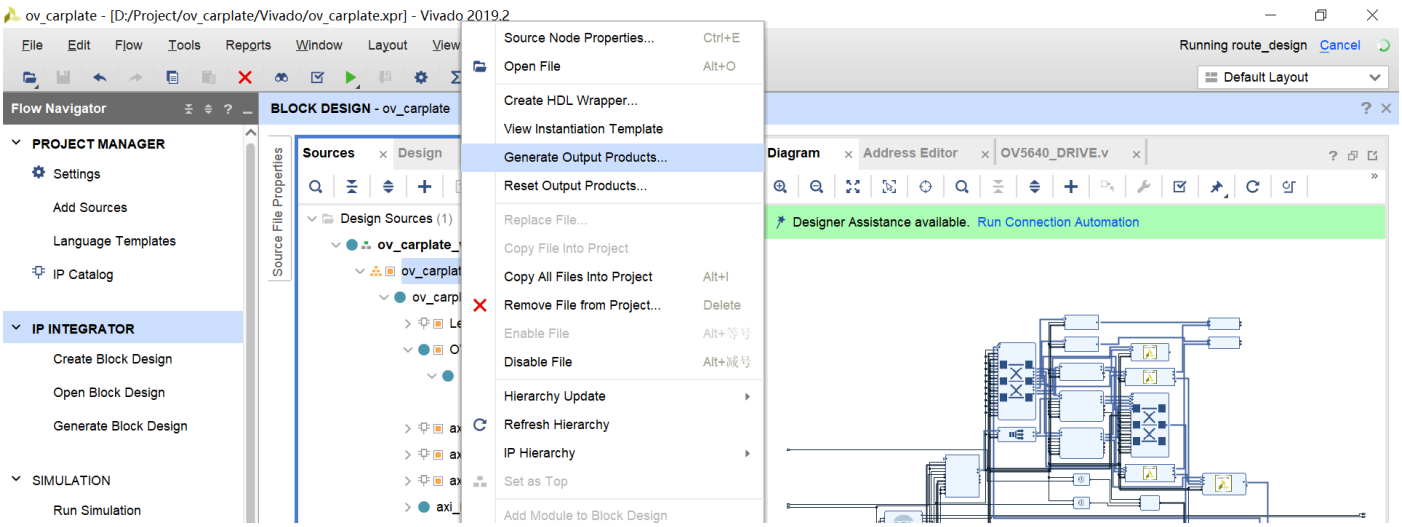
2 添加 IP 目录



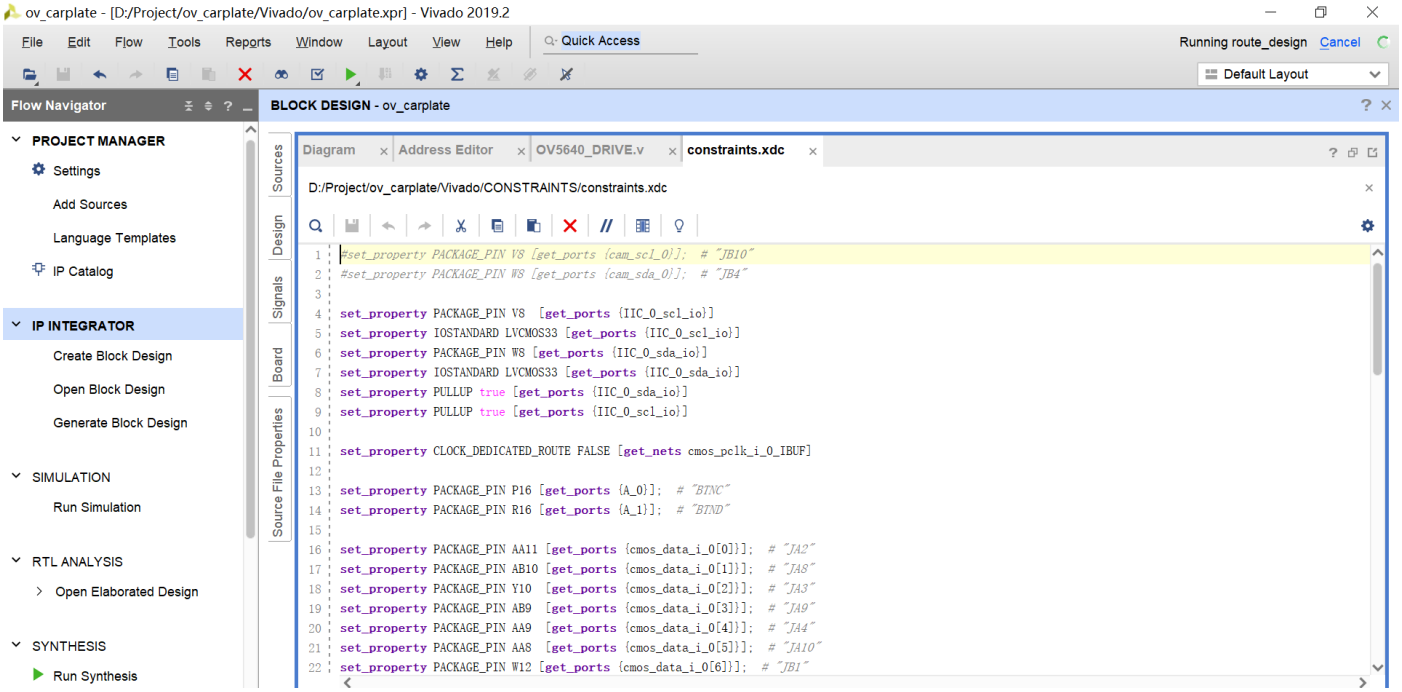
3 构建 Block Design



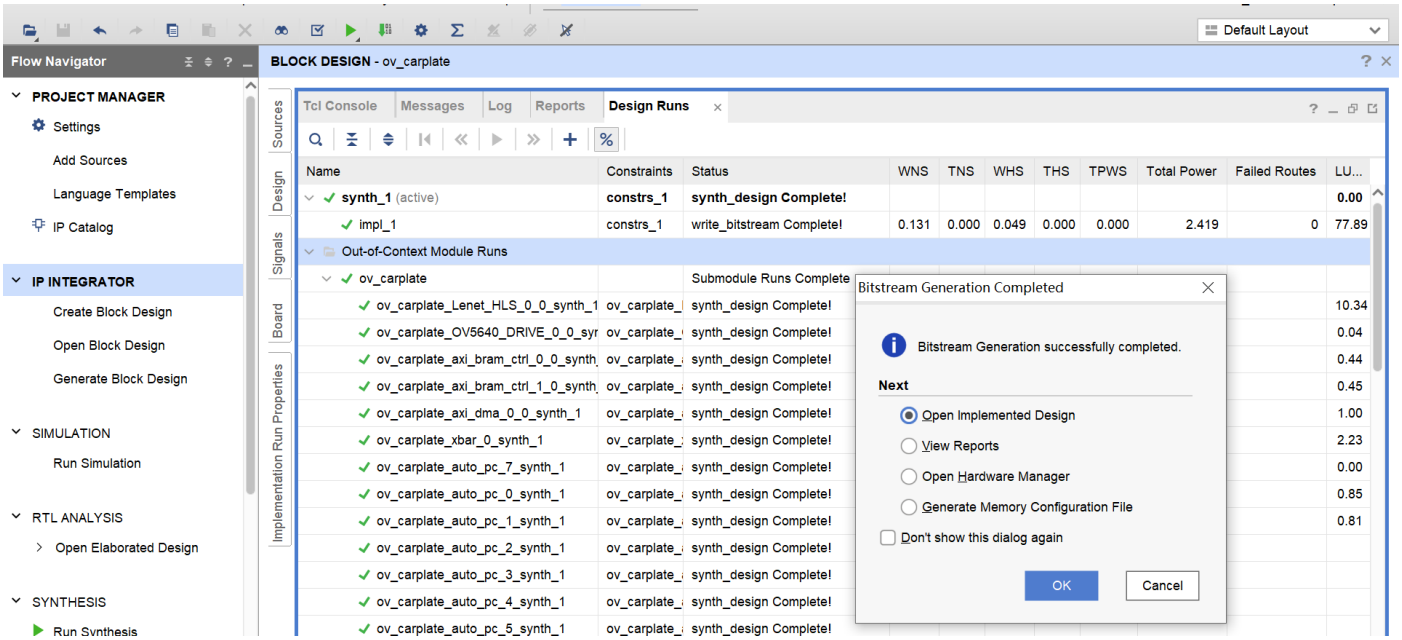
4 Generate Output Product，Create HDL Wrapper



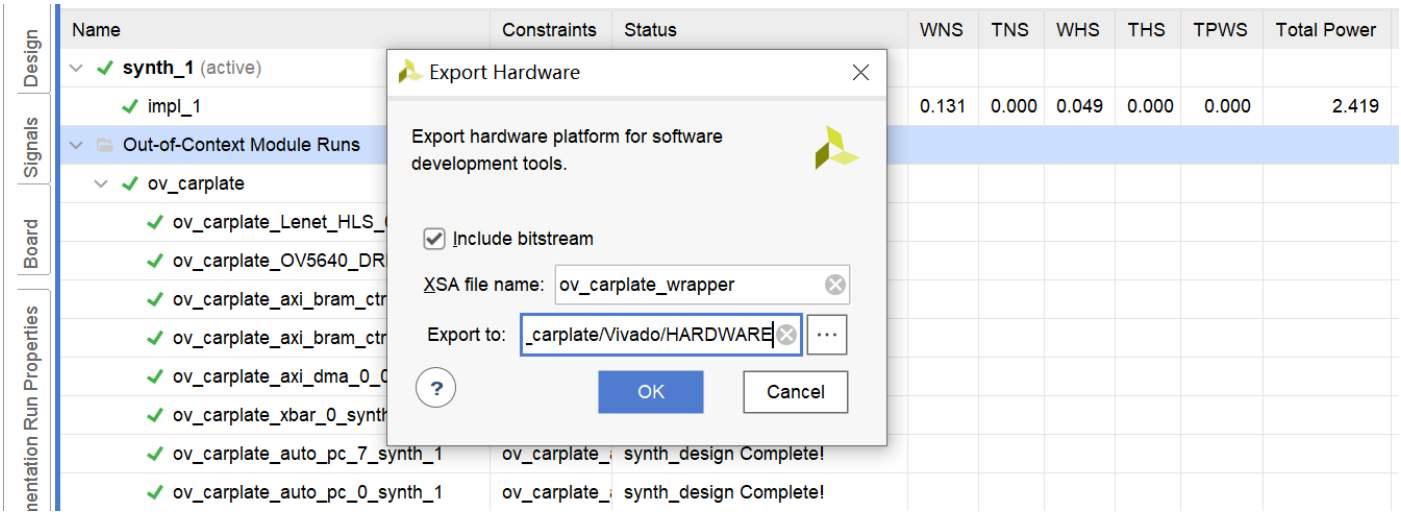
5 编写约束文件



6 Synthesis，Implementation，Generate Bitstream

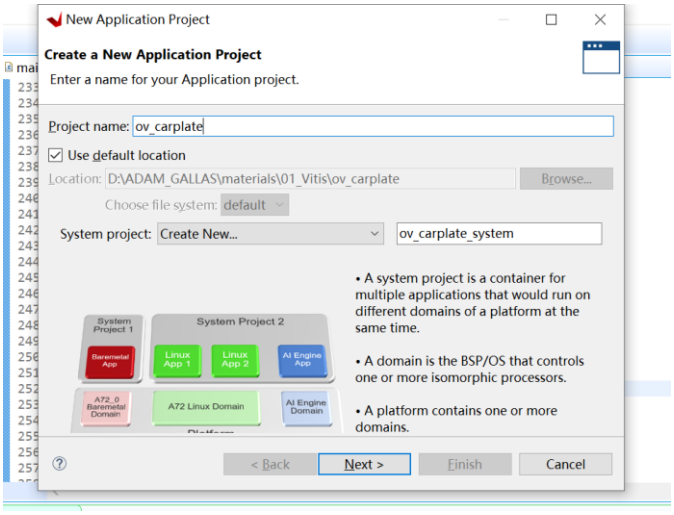


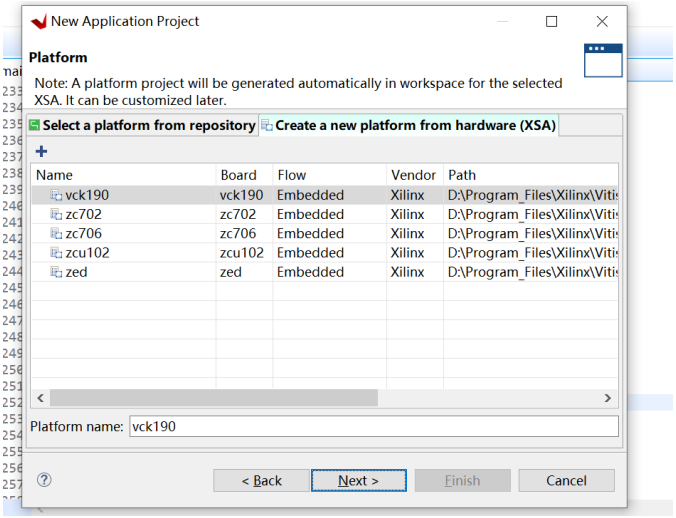
7 Export Hardware



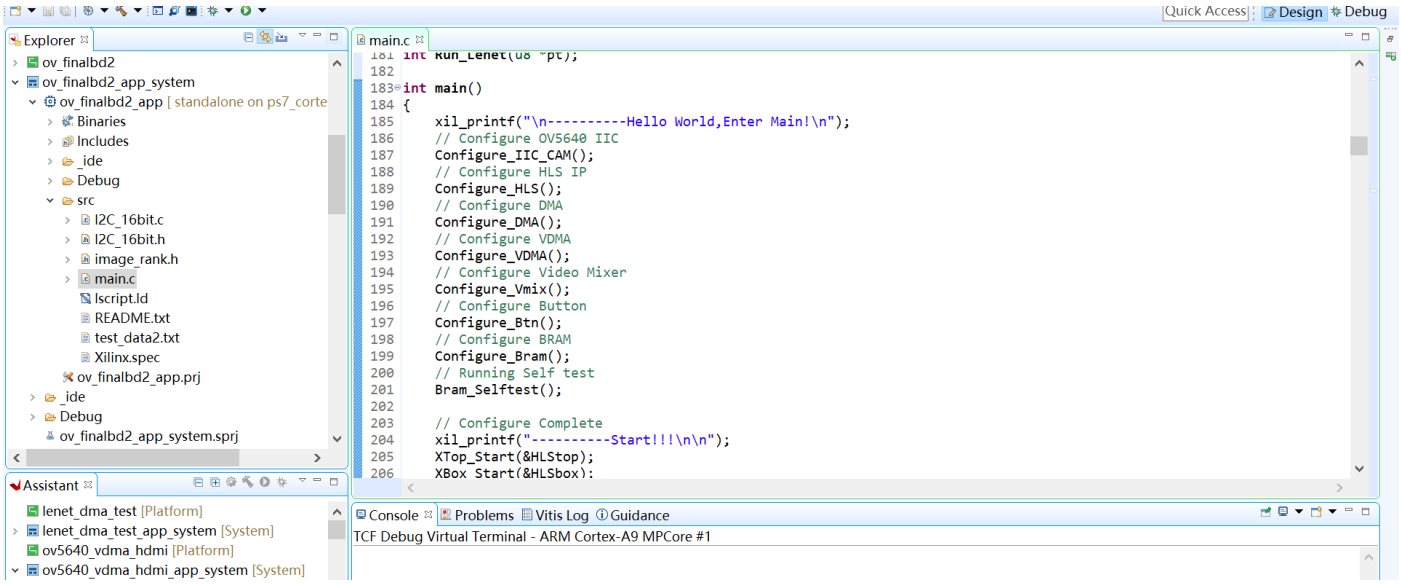
步骤 6 使用 Vitis 进行 PS 端与 IP 配置

1 新建工程

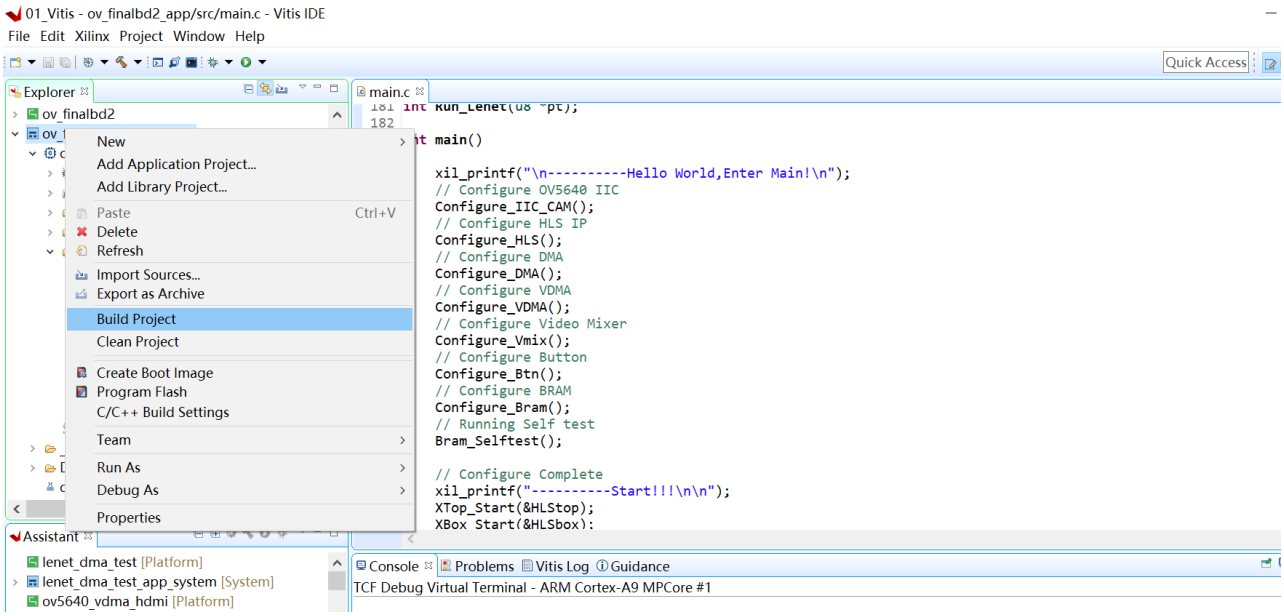




2 初始化 IP



3 Build Project，进行初步测试



步骤 7 PS 端图像处理算法设计

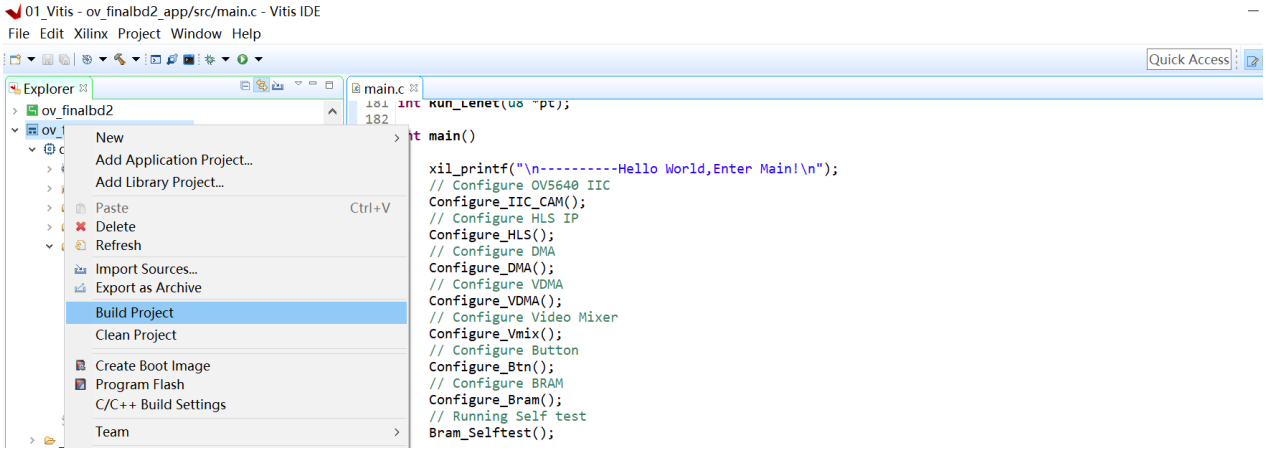
1 编写子过程函数

```
169 // BFS
170 int Bfs_Bounding_Edge();
171 // Cut plate get single char
172 void Get_Plate_Pic();
173 int Get_UpLow_Boundary();
174 void Get_Project_Array();
175 void Get_Characters_Edge();
176 void Get_Single_Char();
177 void Max_Interval_Resize(u8 *src,u8 *dst,int src_r,int src_c,int dst_r,int dst_c);
178 u8 Otsu_Implement(u8 *src,int src_r,int src_c);
179 void Threshold_Binary(u8 *src,u8 threshold,int src_r,int src_c);
```

2 进一步整合拼接

```
225
226   while (1){
227       status=Bfs_Bounding_Edge();
228       if(status==0){
229           continue;
230       }
231       Get_Plate_Pic();
232       thre=Otsu_Implement(plate_pic,PLATE_ROWS,PLATE_COLS);
233       Threshold_Binary(plate_pic,thre,PLATE_ROWS,PLATE_COLS);
234
235       status=Get_UpLow_Boundary();
236       if(status==0){
237           continue;
238       }
239       Get_Project_Array();
240       Get_Characters_Edge();
241       Get_Single_Char();
242       Xil_DCacheDisable();
243       recognition_result[0]=Run_Lenet(char1_r);
244       recognition_result[1]=Run_Lenet(char2_r);
```

3 Build Project



4 连接串口，上板验证测试