Ascend C实践与实验报告

认证成绩



证书

初级:



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中级:

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实验过程

1. 代码编写

WXXXXXX

• op_host: sinh_custom.cpp

WXXXXX

```
#include "sinh custom tiling.h"
#include "register/op def registry.h"
namespace optiling {
/**
Tiling Func负责对输入数据进行分块(Tile)处理。分块处理的好处在于,可以并行计算不同块中的数据,提升计算效率。
BLOCK DIM 定义了每次计算操作需要处理的块的数量。
TILE NUM 定义了在每个计算块中进一步将数据划分为更小的子块。每个子块的数据大小由blocklength/TILE NUM来决定。
该方法将 totalLength 和 TILE_NUM 此类方法保存在tiling对象中,随后将这些信息写入`RawTilingData`中
**/
static ge::graphStatus TilingFunc(gert::TilingContext* context)
   SinhCustomTilingData tiling;
   //考生自行填充
   const uint32 t BLOCK DIM = 8;
   const uint32 t TILE NUM = 8;
   uint32 t totalLength = context->GetInputShape(0)->GetOriginShape().GetShapeSize();
   context->SetBlockDim(BLOCK_DIM);
   tiling.set totalLength(totalLength);
   tiling.set_tileNum(TILE_NUM);
   tiling.SaveToBuffer(context->GetRawTilingData()->GetData(),
   context->GetRawTilingData()->GetCapacity());
   context->GetRawTilingData()->SetDataSize(tiling.GetDataSize());
   size t *currentWorkspace = context->GetWorkspaceSizes(1);
   currentWorkspace[0] = 0;
   return ge::GRAPH SUCCESS;
/**
这个函数定义了输入与输出的形状推理逻辑、保证输入和输出的形状是相同的。
**/
```

```
namespace ge {
static ge::graphStatus InferShape(gert::InferShapeContext* context)
{
   const gert::Shape* x1_shape = context->GetInputShape(0);
   gert::Shape* y shape = context->GetOutputShape(0);
   *y_shape = *x1_shape;
   return GRAPH SUCCESS;
该类定义了一个自定义的sinh算子,明确了输入和输出的张量格式和数据类型(DT FLOAT16),并且指定该算子的推理形状函数是InferShape,Tiling函数是TilingFu
最后,通过OP ADD(SinhCustom)将该算子注册到Ascend编译器中。
**/
namespace ops {
class SinhCustom : public OpDef {
public:
   explicit SinhCustom(const char* name) : OpDef(name)
       this->Input("x")
           .ParamType(REQUIRED)
           .DataType({ge::DT_FLOAT16})
           .Format({ge::FORMAT_ND})
           .UnknownShapeFormat({ge::FORMAT_ND});
       this->Output("y")
           .ParamType(REQUIRED)
           .DataType({ge::DT_FLOAT16})
           .Format({ge::FORMAT ND})
           .UnknownShapeFormat({ge::FORMAT_ND});
       this->SetInferShape(ge::InferShape);
       this->AICore()
```

```
.SetTiling(optiling::TilingFunc);
       this->AICore().AddConfig("ascend310b");
   }
};
OP_ADD(SinhCustom);
• op_host:sinh_custom_tilling.h
#include "register/tilingdata_base.h"
/**
这里定义了tiling数据结构的字段totalLength和tileNum,它们分别表示输入数据的总长度和分块数目。通过REGISTER_TILING_DATA_CLASS将SinhCustomTilingData
**/
namespace optiling {
BEGIN_TILING_DATA_DEF(SinhCustomTilingData)
 //考生自行定义tiling结构体成员变量
 TILING_DATA_FIELD_DEF(uint32_t, totalLength);
 TILING_DATA_FIELD_DEF(uint32_t, tileNum);
END_TILING_DATA_DEF;
REGISTER_TILING_DATA_CLASS(SinhCustom, SinhCustomTilingData)
}
• op_kernel:sinh_custom.cpp
```

```
#include "kernel operator.h"
using namespace AscendC;
constexpr int32 t BUFFER NUM = 2;
class KernelSinh {
public:
   __aicore__ inline KernelSinh() {}
   /**
   该函数负责初始化全局和局部缓存、块和Tile的长度,并根据tileNum和blockLength来计算tileLength。
xGm.SetGlobalBuffer 和 yGm.SetGlobalBuffer 初始化全局内存上的输入和输出数据区域。
pipe.InitBuffer 初始化了多个队列和临时缓冲区,用于算子执行过程中数据的缓存和处理。
   **/
    aicore inline void Init(GM ADDR x,GM ADDR y,uint32 t totalLength, uint32 t tileNum)
       //考生补充初始化代码
       ASSERT(GetBlockNum() != 0 && "block dim can not be zero!");
       this->blockLength = totalLength / GetBlockNum();
       this->tileNum = tileNum;
       ASSERT(tileNum != 0 && "tile num can not be zero!");
       this->tileLength = this->blockLength / tileNum / BUFFER NUM;
       xGm.SetGlobalBuffer(( gm DTYPE X *)x + this->blockLength * GetBlockIdx(),
       this->blockLength);
       yGm.SetGlobalBuffer((__gm__ DTYPE_Y *)y + this->blockLength * GetBlockIdx(),
       this->blockLength);
       pipe.InitBuffer(inQueueX, BUFFER NUM, this->tileLength * sizeof(DTYPE X));
       pipe.InitBuffer(outQueueY, BUFFER NUM, this->tileLength * sizeof(DTYPE Y));
       pipe.InitBuffer(tmpBuffer1, this->tileLength * sizeof(DTYPE_X));
       pipe.InitBuffer(tmpBuffer2, this->tileLength * sizeof(DTYPE X));
       pipe.InitBuffer(tmpBuffer3, this->tileLength * sizeof(DTYPE_X));
       pipe.InitBuffer(tmpBuffer4, this->tileLength * sizeof(DTYPE X));
   }
```

```
aicore inline void Process()
   {
       /*
       Process函数执行主循环,每次循环中执行三个步骤:从全局内存拷贝数据到局部内存(CopyIn),计算(Compute),然后将结果从局部内存拷贝回全局内存
       */
       int32 t loopCount = this->tileNum*BUFFER NUM;
       for (int32 t i = 0; i < loopCount; i++) {
           CopyIn(i);
           Compute(i);
           CopyOut(i);
       }
   }
private:
   __aicore__ inline void CopyIn(int32_t progress)
   {
       //考生补充算子代码
       LocalTensor<DTYPE X> xLocal = inQueueX.AllocTensor<DTYPE X>();
       DataCopy(xLocal, xGm[progress * this->tileLength], this->tileLength);
       inQueueX.EnQue(xLocal);
    __aicore__ inline void Compute(int32_t progress)
       //考生补充算子计算代码
       LocalTensor<DTYPE X> xLocal = inQueueX.DeQue<DTYPE X>();
       LocalTensor<DTYPE Y> yLocal = outQueueY.AllocTensor<DTYPE Y>();
       LocalTensor<DTYPE X> tmpTensor1 = tmpBuffer1.Get<DTYPE X>();
       LocalTensor<DTYPE X> tmpTensor2 = tmpBuffer2.Get<DTYPE X>();
       LocalTensor<DTYPE X> tmpTensor3 = tmpBuffer3.Get<DTYPE X>();
       LocalTensor<DTYPE X> tmpTensor4 = tmpBuffer4.Get<DTYPE X>();
       DTYPE X inputVal1 = -1;
       DTYPE X inputVal2 = 0.5;
```

```
//sinh(x) = (exp(x) - exp(-x)) / 2.0
       /**
       将输入张量乘以-1(Muls),得到-x。
               计算exp(-x)(Exp)。
               计算exp(x)。
               计算exp(x) - exp(-x)(Sub)。
               将结果乘以0.5,得到sinh(x)的结果(Muls)。
       **/
       Muls(tmpTensor1, xLocal, inputVal1, this->tileLength);
       Exp(tmpTensor2, tmpTensor1, this->tileLength);
       Exp(tmpTensor3, xLocal, this->tileLength);
       Sub(tmpTensor4, tmpTensor3, tmpTensor2, this->tileLength);
       Muls(yLocal, tmpTensor4, inputVal2, this->tileLength);
       outQueueY.EnQue<DTYPE Y>(yLocal);
       inQueueX.FreeTensor(xLocal);
   }
   __aicore__ inline void CopyOut(int32_t progress)
       //考生补充算子代码
       LocalTensor<DTYPE_Y> yLocal = outQueueY.DeQue<DTYPE_Y>();
       DataCopy(yGm[progress * this->tileLength], yLocal, this->tileLength);
       outQueueY.FreeTensor(yLocal);
   }
private:
   TPipe pipe;
   //create queue for input, in this case depth is equal to buffer num
   TQue<QuePosition::VECIN, BUFFER NUM> inQueueX;
   //create queue for output, in this case depth is equal to buffer num
   TQue<QuePosition::VECOUT, BUFFER NUM> outQueueY;
   GlobalTensor<half> xGm;
   GlobalTensor<half> yGm;
```

```
//考生补充自定义成员变量
   TBuf<QuePosition::VECCALC> tmpBuffer1, tmpBuffer2, tmpBuffer3, tmpBuffer4;
   uint32_t blockLength;
   uint32_t tileNum;
   uint32_t tileLength;
};
这是最终的自定义内核函数,通过Init函数初始化操作,并调用Process函数执行具体计算。
**/
extern "C" global aicore void sinh custom(GM ADDR x, GM ADDR y, GM ADDR workspace, GM ADDR tiling) {
   GET_TILING_DATA(tiling_data, tiling);
   KernelSinh op;
   //补充init和process函数调用内容
   op.Init(x, y, tiling_data.totalLength, tiling_data.tileNum);
   op.Process();
2. cd /root/SinhCustom/SinhCustom, 运行bash build.sh
  成功后出现build out文件夹
3. cd /root/SinhCustom/SinhCustom/build_out 尝试运行./custom_opp_ubuntu_aarch64.run
```

- success.
- 4. cd /root/SinhCustom/AclNNInvocation bash run.sh
- 报错, 找不到相关.h文件

于是将SinhCustom\SinhCustom\build out\autogen中的.h文件和.cpp文件复制到SinhCustom11\SinhCustom\AclNNInvocation中include和scr文 件夹中,然后在cmakelist中添加对应的编译选项。

- 重新运行,依旧报错,找不到相关的库,执行命令,找到对应库的位置,然后在cmakelist里链接相关的库。
- 重新运行,成功完成实验。