

Creating Custom Layers in Tensorflow

Matt Angus

University of Waterloo

m2angus@gsd.uwaterloo.ca

July 5, 2017

Introduction

Some Terminology

- Host - code/memory
- Device - code/memory

General Steps

- ① Copy data from host memory to device memory
- ② Load program onto device, and execute
- ③ Copy Results from device memory to host memory

Writing and Calling CUDA Code

```
--global__ void cudaKernel(){  
    //do something on the gpu  
}  
  
int main() {  
    cudaKernel<<<1,1>>>();  
}
```

- `-- global__` marks device code
- `<<<>>>` marks a call from host to device

A word on pointers

There are two types of pointers, host pointers and device pointers. Device pointers can be created manipulated with

- `cudaMalloc()`
- `cudaFree()`
- `cudaMemcpy()`

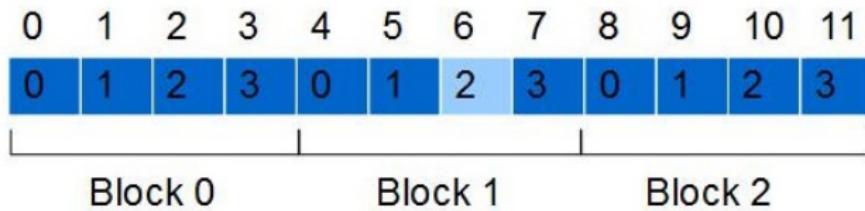
Analogous to C equivalents `malloc()`, `free()`, `memcpy()`.

Passing data between host and device

```
--global__ void doWork(int* a){  
    //a points to a memory address on the device  
}  
  
int main() {  
    int a = 10;  int *d_a;  
    int size = sizeof(int);  
  
    cudaMalloc((void**)&d_a, size);  
    cudaMemcpy(d_a, &a, size, cudaMemcpyHostToDevice)  
    doWork<<<1,1>>>(d_a);  
    cudaMemcpy(&a, d_a, size, cudaMemcpyDeviceToHost)  
  
    cudaFree(d_a);  
    return 0;  
}
```

Making it Parallel

- Call `doWork<<<N,M>>>(d_a);`
- N blocks and M threads per block
- Use `blockIdx.x` and `threadIdx.x` to split up work



- `index = threadIdx.x + blockIdx.x * M`
- `index = 2 + 1 * 4 = 6`

Making it Parallel

```
--global--  
void doWork(int* a, int* b, int* c, int N);  
  
void launchKernel(int* a, int* b, int* c, int N) {  
    int *d_a; int* d_b; int* d_c;  
    int size = sizeof(int)*N;  
  
    cudaMalloc((void**)&d_a, size);  
    cudaMalloc((void**)&d_b, size);  
    cudaMalloc((void**)&d_c, size);  
    cudaMemcpy(d_a, &a, size, cudaMemcpyHostToDevice)  
    cudaMemcpy(d_b, &b, size, cudaMemcpyHostToDevice)
```

Making it Parallel

```
const int t_per_b = 512; //multiple of 32
doWork<<<(N + t_per_b-1) / t_per_b,
            t_per_b>>>(d_a, d_b, d_c, N);
cudaMemcpy(&c, d_c, size, cudaMemcpyDeviceToHost)

cudaFree(d_a); cudaFree(d_b); cudaFree(d_c);
}

__global__
void doWork(int* a, int* b, int* c, int N){
    int i = threadIdx.x + blockIdx.x * M;
    if(i < N)
        c[i] = a[i] + b[i];
}
```

Some Things to Note

- Does not handle shared access
- `doWork<<<A, B, C, D>>>()`
 - A: # Blocks
 - B: # Threads/Block
 - C: Amount of shared memory
 - D: Device stream

Tensorflow Overview

- Data is stored in tensors (nd-arrays)
- Build a computation graph with operations on tensors
- flow tensors through operations

The Bad News

- Don't know how to do something?
 - start here: <https://fossies.org/dox/tensorflow-1.2.0/>
 - better get digging (through code)

InferenceContext member functions

Status	Merge (ShapeHandle <i>in0</i> , ShapeHandle <i>in1</i> , ShapeHandle * <i>out</i>) TF_MUST_USE_RESULT
Status	MergePrefix (ShapeHandle <i>s</i> , ShapeHandle <i>prefix</i> , ShapeHandle * <i>s_out</i> , ShapeHandle * <i>prefix_out</i>) TF_MUST_USE_RESULT
Status	Merge (DimensionHandle <i>d0</i> , DimensionHandle <i>d1</i> , DimensionHandle * <i>out</i>) TF_MUST_USE_RESULT
Status	Subshape (ShapeHandle <i>s</i> , int64 <i>start</i> , ShapeHandle * <i>out</i>) TF_MUST_USE_RESULT
Status	Subshape (ShapeHandle <i>s</i> , int64 <i>start</i> , int64 <i>end</i> , ShapeHandle * <i>out</i>) TF_MUST_USE_RESULT

Overview of Steps

- Register the C++ operation. Define inputs, outputs, and types.
- Implement the kernel for a specific device (GPU, CPU)
- Create a python wrapper
- Implement Gradient (if we have time)

Useful Tensorflow Classes

- **OpKernel**
 - Main op class to inherit from
- **Tensor**
 - Wrapper for data with useful functions
- **OpKernelContext**
 - Context for op when computation begins
 - Contains input, allocates output
- **InferenceContext**
 - Used to infer and validate shape of input and output
 - Used when building the graph
- **Status**
 - communicate to and from TF if everything is okay

Useful Tensorflow Macros

- `TF_RETURN_IF_ERROR(stmnt)`
 - Catches any errors thrown by running `stmtnt`
 - returns a bad Status with the correct message
- `OP_REQUIRES_OK(ctx, stmtnt)`
 - execute `stmtnt`
 - ensure that the status of `ctx` is "ok".
- `OP_REQUIRES(ctx, boolean, err)`
 - if `boolean == false` throw `err`
- `CUDA_1D_KERNEL_LOOP(ind, N)`
 - handles indexing into the 1d array
 - calculates `ind` based on `blockIdx` etc.

Let's Get Coding!

```
git@github.com:mattangus/TF-Custom-Op-Workshop.git
git fetch --all
git checkout skeleton
```

Registering the Op

Define interface between for operation. This indicates to TF what it should expect when building the graph.

```
REGISTER_OP("CustomAdd")
    .Input("a:<T>")
    .Input("b:<T>")
    .Output("c:<T>")
    .Attr("T:<{int32, float32, float64}")
    .SetShapeFn(ShapeFn);
```

Registering the Op

Instantiate the kernel with a type.

```
REGISTER_KERNEL_BUILDER(  
    Name("CustomAdd")  
    .Device(DEVICE_CPU)  
    .TypeConstraint<double>("T"),  
    CustomAddOp<double>);
```

Shape Inference

```
ShapeHandle a_shape;
TF_RETURN_IF_ERROR(c->WithRank(c->input(0),
    4, &a_shape));

ShapeHandle b_shape;
TF_RETURN_IF_ERROR(c->WithRank(c->input(1),
    4, &b_shape));
```

Shape Inference

```
for(int i = 0; i < 4; i++)
{
    DimensionHandle a_dim = c->Dim(a_shape,i);
    DimensionHandle b_dim = c->Dim(b_shape,i);
    if (c->Value(a_dim) != c->Value(b_dim))
        return errors::InvalidArgument(
            "a and b dim mismatch");
}

c->set_output(0, c->input(0));
```

Allocating Output

```
Tensor* output_tensor = nullptr;
OP_REQUIRES_OK(context,
    context->allocate_output(0,
    a_tensor.shape(),&output_tensor));

auto output = output_tensor->flat<dtype>();
```

Add the Tensors

```
const int N = output.size();
const dtype* a = a_flat.data();
const dtype* b = b_flat.data();
dtype* c = output.data();
for(int i = 0; i < N; i++)
{
    c[i] = a[i] + b[i];
}
```

Changing to the GPU

DEVICE_CPU > DEVICE_GPU

```
launchAddKernel<dtype>(a_flat.data(),
    b_flat.data(), output.data(), N);
```

```
template void launchAddKernel<double>(
    const double* a, const double* b,
    double* c, int N);
```

Changing to the GPU

```
CUDA_1D_KERNEL_LOOP(index, N)
{
    c[index] = a[index] + b[index];
}

AddKernel<dtype><<<
(N + kThreadsPerBlock - 1) / kThreadsPerBlock,
kThreadsPerBlock>>>(a, b, c, N);
```

Leverage the Template

```
#define REGISTER_KERNEL(type) \
REGISTER_KERNEL_BUILDER( \
Name("CustomAdd") \
.Device(DEVICE_GPU) \
.TypeConstraint<type>("T"), \
CustomAddOp<type>) \
 \
REGISTER_KERNEL(int); \
REGISTER_KERNEL(float); \
REGISTER_KERNEL(double); \
 \
#undef REGISTER_KERNEL
```

Leverage the Template

```
#define ADD_KERNEL_TYPE(type) \
template void launchAddKernel<type>( \
    const type* a, const type* b, \
    type* c, int N) \
\
ADD_KERNEL_TYPE(int); \
ADD_KERNEL_TYPE(float); \
ADD_KERNEL_TYPE(double); \
\
#undef ADD_KERNEL_TYPE
```

Questions?