HLS for Human Activity Recognition using CNN in Keras

Group Number: 2

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1. Description of the model:

Task of model	Number of layers	Type of Lavers		Model size
Human Activity Recognition		Input Layer	1	
	7	Conv2D	1	
		MaxPooling2D	1	8.9 MB
		Reshape	1	
		Dense Layer	3	

Topology:

Layer name	Layer type	Input shape	Output shape
Conv2d_1_input	InputLayer	[[None,90,3,1]]	[None,90,3,1]
Conv2d_1	Conv2D	[[None,90,3,1]]	[None,89,2,128]
max_pooling2d_1	MaxPooling2D	[[None,89,2,128]]	[None,44,1,128]
faltten_1	Reshape	[[None,44,1,128]]	[None,5632]
dense_1	Dense	[[None,5632]]	[None,128]
dense_2	Dense	[[None,128]]	[None,128]
dense_3	Dense	[[None,128]]	[None,6]

2. Changes made to make keras2c generated files synthesizable and description of the change made.

Simulation and Synthesis:

 Resolved linkage errors: Various functions defined in include folder of keras2c in our converted code. So, we included those headers for resolving linkage errors.

```
#include "k2c_include.h"
#include "k2c_tensor_include.h"
#include "k2c_pooling_layers.h"
#include "k2c_convolution_layers.h"
#include "k2c_core_layers.h"
#include "k2c_helper_functions.h"
#include "k2c_recurrent_layers.h"
#include "k2c_activations.h"
```

- Resolved redefinition errors: In multiple 'for' loops, we were getting redefinition errors due to same definition of loop iterators in each loop.
 So, we declared that variable once and update for each loop respectively.
- Changed dynamic to static memory allocation for 'array' data member of all k2c_tensor structures to resolve "Unknown size at compile time" error. We declared the array with a size of 725000.

```
#define MAX_TENSOR_ARRAY_SIZE 725000

/**
    * tensor type for keras2c.
    */
struct k2c_tensor
{
        /** Pointer to array of tensor values flattened in row major order. */
        //float * array;
        float array[MAX_TENSOR_ARRAY_SIZE];

        /** Rank of the tensor (number of dimensions). */
        size_t ndim;

        /** Number of elements in the tensor. */
        size_t numel;

        /** Array, size of the tensor in each dimension. */
        size_t shape[K2C_MAX_NDIM];
};
```

• Removed memcpy() functions and replaced them with a 'for' loop.

```
// memset(C, 0, outrows*outcols*sizeof(C[0]));
size_t i;

for (i = 0; i < outrows * outcols; ++i) {
    C[i] = 0;
}</pre>
```

- Removed function pointers by directly calling the function they were pointing to.
- Global declaration of all big arrays and declarations of k2c_tensor structures to solve segmentation fault that was coming due to insufficient stack memory.

Co-simulation:

• Changed source file and testbench files' extension from '.c' to '.cpp' files to resolve the below error:

```
INFO: [HLS 200-10] In directory '/home/akshat/original_cnn_struct/original_cnn_struct/solution1/sim/wrapc'
clang: warning: argument unused during compilation: '-fno-builtin-isinf'
clang: warning: argument unused during compilation: '-fno-builtin-isinf'
In file included from /home/akshat/original_cnn_struct/converted.c:1:
In file included from /usr/include/math.h:43:
/usr/include/x86_64-linux-gnu/bits/floatn.h:74:52: error: unsupported machine mode '__TC__'
typedef _Complex float _cfloat128 _attribute_ ((_mode_ (_TC__)));
/usr/include/x86_64-linux-gnu/bits/floatn.h:86:9: error: unknown type name '__float128'; did you mean '__cfloat128'?
typedef _float128 _Float128;
/usr/include/x86_64-linux-gnu/bits/floatn.h:74:24: note: '__cfloat128' declared here
typedef _Complex float _cfloat128 _attribute_ ((_mode_ (_TC__)));
2 errors generated.
```

3. Changes made to generate HLS4ML report.

Commented the below line in 'build_prj.tcl' to solve the mentioned error:

Line commented	config_schedule -enable_dsp_full_reg=false
Error	ERROR: [HLS 200-101] 'config_schedule': Unknown option '-enable_dsp_full_reg=false'.:

Following pragmas were removed from 'nnet_dense_latency.h' to solve 'Stop unrolling loop Product-1' error due to pipelining:

#pragma HLS function_instantiate variable=weights,biases
#pragma HLS PIPELINE II=CONFIG_T::reuse_factor
#pragma HLS ARRAY_PARTITION variable=biases complete
#pragma HLS ARRAY_PARTITION variable=mult complete
#pragma HLS ARRAY_PARTITION variable=acc complete
#pragma HLS ALLOCATION operation instances=mul
limit=CONFIG_T::multiplier_limit

Following pragmas were removed from pooling2d_cl () function in 'nnet_pooling.h' to solve 'Stop unrolling loop Loop-1' error :

#pragma HLS PIPELINE II=CONFIG_T::reuse_factor
<pre>#pragma HLS PIPELINE II=CONFIG_T::reuse_factor</pre>
#pragma HLS ARRAY_PARTITION variable=pool complete dim=0

Following pragmas were removed from 'nnet_activation.h' to solve 'Excessive memory usage' error :

#pragma HLS PIPELINE
#pragma HLS UNROLL

#pragma HLS INLINE

Following pragmas in 'myproject.cpp' were modified:

Original	Modified
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer2_out complete dim=0	variable=layer2_out block factor=8
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer3_out complete dim=0	variable=layer3_out block factor=8
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer4_out complete dim=0	variable=layer4_out block factor=8
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer6_out complete dim=0	variable=layer6_out block factor=8
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer8_out complete dim=0	variable=layer8_out block factor=8
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer9_out complete dim=0	variable=layer9_out block factor=8
#pragma HLS ARRAY_PARTITION	#pragma HLS ARRAY_PARTITION
variable=layer10_out complete	variable=layer10_out block factor=4
dim=0	

These were previously partitioned with type=complete. Due to this, each element of every array was stored in a register which was increasing resource usage and we were getting "Excessive memory usage" error. All arrays except layer10_out was partitioned with type=block and factor=8. The array 'layer10_out' was partitioned with type=block and factor=4 as this array's size is less than 8.

4. Depedencies and versions required:

Dependencies

- 1. Tensorflow version: 2.15.1
- 2. Pydot
- 3. Graphviz
- 4. Python: Python 3.10.x is needed
- 5. Hls4ml

Commands to install above dependencies

- 1. pip install tensorflow==2.15.x
- 2. pip install pydot
- 3. pip install graphviz
- 4. sudo apt install python3
- 5. pip install hls4ml

5. **Optimizations**:

* Resource optimization:

The maximum size of 'array' data member of k2c_tensor is 720896. Most of the arrays are of size less than 20K. This results in a lot of unused memory.

For optimizing resource, we removed the k2c_tensor structures completely and wherever we were passing the structure (say) A, we modified that to pass all the 4 parameters of structure : A_array, A_ndim, A_numel, A_shape as pointers.

```
//void k2c_dense_func2(k2c_tensor* output, const k2c_tensor* input, const k2c_tensor2* kernel,

// const k2c_tensor* bias, int idx, float * fwork) {

void k2c_dense_func2(float* output_array, size_t* output_ndim, size_t* output_numel, size_t* output_shape,

const float* input_array, const size_t* input_array, const size_t* input_numel, const size_t* input_shape,

const float* kernel_array, const size_t* kernel_ndim, const size_t* kernel_numel, const size_t* kernel_shape,

const float* bias_array, const size_t* bias_ndim, const size_t* bias_numel, const size_t* bias_shape,

int idx, float * fwork) {
```

Latency optimization:

For improving latency, we have used array partitioning and loop pipelining. For reference, we have given the snapshot of k2c_affine_matmul() function:

```
// original
 void k2c affine matmul(float * C, const float * A, const float * B, const float * d,
                           const size t outrows,const size t outcols, const size t innerdim) {
     // make sure output is empty
     size t i;
      size t j;
    // memset(C, 0, outrows*outcols*sizeof(C[0]));
      for (int k = 0; k < outrows; ++k) {</pre>
               for (int l = 0; l < outcols; ++l) {</pre>
                    C[k * outcols + l] = 0.0f;
     for (i= 0 ; i < outrows; ++i) {</pre>
          const size t outrowidx = i*outcols;
          const size t inneridx = i*innerdim;
          for (j = 0; j < outcols; ++j) {</pre>
              for (size t k = 0; k < innerdim; ++k) {
                   C[outrowidx+j] += A[inneridx+k] * B[k*outcols+j];
              C[outrowidx+j] += d[j];
          }
     }
}
// After optimization
void k2c affine matmul(float * C, const float * A, const float * B, const float * d,
                      const size_t outrows,const size_t outcols, const size_t innerdim) {
   #pragma HLS ARRAY_PARTITION variable=B cyclic factor=8
     for (size t k = 0; k < outrows; ++k) {
        const size_t outrowidx = k * outcols;
        const size t inneridx = k * innerdim;
        for (size t l = 0; l < outcols; l+=8) {</pre>
               #pragma HLS pipeline II=1
            float sum[8]= {0};
            for(size t i = 0; i < innerdim; i++)</pre>
                float a val = A[inneridx + i];
                #pragma HLS UNROLL factor = 8
                for(size t ll=0; ll<8; ll++)</pre>
                    sum[ll] += a val * B[i * outcols + l + ll];
            }
            for(size_t ll=0; ll<8; ll++)</pre>
                C[outrowidx + l + ll] = sum[ll] + d[l + ll];
            }
       }
}
```

6. Results:

• Latency and area overhead table for Baseline (Unoptimized).

─ Summary				
Name	BRAM_18K	DSP48E	FF	LUT
DSP	-	-	-	-
Expression	-	-	0	626
FIFO	-	-	-	-
Instance	64	167	38730	29763
Memory	30783	-	1056	8
Multiplexer	-	-	-	1900
Register	-	-	1185	-
Total	30847	167	40971	32297
Available	730	740	269200	129000
Utilization (%)	4225	22	15	25

Result							
			Latency		In	ter	/al
RTL	Status	min	avg	max	min	avg	max
VHDL	NA	NA	NA	NA	NA	NΑ	NA
Verilog	Pass	29607917	29607917	29607917	NA	NΑ	NA

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• HLS4ML generated Latency and area overhead table.

== Utilization Estimates ______ * Summary: +----+ +----+ +----+ | 1955| 400| 151624| 108606| +----+ |Available | 730| 740| 269200| 129000| +----+ |Utilization (%) | 267| 54| 56| 84| +-----+ + Latency (clock cycles): * Summary: +-----+ | Latency | Interval | Pipeline | min | max | min | max | Type | +-----+ 9273017 | 9746106 | 8685445 | 8685445 | dataflow | +----+

• Latency and area overhead table for resource optimized solution.

-				
Name	BRAM_18K	DSP48E	FF	LUT
DSP	-	-	-	-
Expression	-	-	0	306
FIFO	-	-	-	-
Instance	2086	34	6803	7064
Memory	114	-	0	0
Multiplexer	-	-	-	623
Register	-	-	333	-
Total	2200	34	7136	7993
Available	730	740	269200	129000
Utilization (%)	301	4	2	6

Result							
			Latency		In	ter	/al
RTL	Status	min	avg	max	min	avg	max
VHDL	NA	NA	NA	NA	NA	NA	NA
Verilog	Pass	16231732	16231732	16231732	NA	NA	NA

• Latency and area overhead table for latency optimized solution.

Summary				
Name	BRAM_18K	DSP48E	FF	LUT
DSP	-	-	-	-
Expression	-	-	0	306
FIFO	-	-	-	-
Instance	2148	139	21364	16557
Memory	116	-	0	0
Multiplexer	-	-	-	1403
Register	-	-	851	-
Total	2264	139	22215	18266
Available	730	740	269200	129000
Utilization (%)	310	18	8	14

Result													
		Latency				Interval							
RTL	Status	min	avg	max	min	avg	max						
VHDL	NA	NA	NA	NA	NA	NΑ	NA						
Verilog	Pass	4524140	4524140	4524140	NA	NΑ	NA						

Comparison:

	LUT	FF	DSP	BRAM	Latency		Clock
Design					Min	Max	period(in ns)
Baseline	32997	40971	167	30847	29607917	29607917	5
HLS4ML	108606	151624	400	1955	9273017	9746106	5
Resource optimized	7993	7136	34	2200	16231732	16231732	5
Latency optimized	18266	22215	139	2264	4524140	4524140	5