FP16 Tests Experiment Summary

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1 Experiment Setup

1.1 Sources of fp16 configs

All fp16 configs are extracted from /mlir/test/mlir-miopen-driver/ that are used for E2E test.

Here is a list of all files that contain at least one f16 config:

- auto_e2e/padding_kernel_gemmK.mlir (3)
- auto_e2e/padding_kernel_gemmN.mlir(2)
- auto_e2e/conv2d_host_validation_f16_fwd.mlir (52, 1 is FIXME)
- auto_e2e/conv2d_host_validation_f16_bwd.mlir (18)
- auto_e2e/padding_kernel_all_fwd.mlir (2)
- $auto_e2e/padding_kernel_all_wrw.mlir$ (2)
- auto_e2e/conv2d_host_validation_f32_fwd.mlir (1, f32 file has a f16 test?)
- e2e_for_pr/Resnet50_validation_nchw.mlir (4)
- e2e_for_pr/Resnet50_validation_nhwc.mlir (4)

- misc_e2e/conv2d_harness_cpu_verifier_kyxc_nhwc_nhwk_f16.mlir (1)
- misc_e2e/conv2d_host_validation_f16.mlir (10, 38 are FIXME)
- misc_e2e/conv2d_host_validation_random.mlir (8)

The number in the parenthesis indicates the number of fp16 configs in the file. There are totally 107 fp16 configs (FIXME configs are not counted).

To test all fp16 configs, each of the above files is broken into unit tests, which contains only one line of RUN and its corresponding CHECKs. This is done by

./splitTest.sh <input file> <dest dir> <data type>

which extracts single tests with -t <data type> from <input test> and put the result unit tests into <dest dir>.

1.2 Settings of fp16 tests

When invoking miopen-gen on each of the config, the following options can be set to control the behavior of the test:

- -rand 1 or -rand 0: choose the seed of the random number generator. 1 means a fixed seed and 0 means a seed that is a function of the time.
- -rand_min min and -rand_max max: choose the range of the random numbers.
- -pv or -pv_with_gpu: choose the validation function method.
- -x2 or : choose whether xdlops is enabled for the GPU kernel.

1.3 Metrics

To verify the results of the GPU kernel, the verification function uses the following metrics to measure the difference between the results of the GPU kernel (kern) and the validation function (val).

1.3.1 Element wise difference

The element wise difference metric computes the maximum difference among all elements that can be written as diff = $\max_{i}(d(\text{val}_i - \text{kern}_i, \text{val}_i))$ When different diff functions (d()) are used, we can have the following element wise difference metrics.

• Absolute difference:

$$\max AbsDiff = \max_{i}(|val_{i} - kern_{i}|)$$

• Relative difference:

$$\text{maxRelDiff} = \max_{i} \left(\frac{|\text{val}_{i} - \text{kern}_{i}|}{|\text{val}_{i}|} \right), |\text{val}_{i}| > 0$$

• Relative difference (ignore small denominators):

$$\text{maxRelDiff}_{\text{old}} = \max_{i} \left(\frac{|\text{val}_{i} - \text{kern}_{i}|}{|\text{val}_{i}|} \right), |\text{val}_{i}| > 1 \times 10^{-3}$$

• Epsilon difference:

$$\text{maxEpsilonDiff} = \max_i \left(\frac{|\text{val}_i - \text{kern}_i|}{\epsilon_{|\text{val}_i|}} \right)$$

where $\epsilon_{|val_i|}$ is the precision of fp16 numbers around $|val_i|$. For more information, check the precision limitations section of Half-precision floating-point format¹.

For debugging purpose, the histograms of the above element wise differences can be printed for an individual test. We will show the histograms of different element wise metrics for the worst test on both MI200 and MI100 in this section.

 $^{^{1} \}rm https://en.wikipedia.org/wiki/Half-precision_floating-point_format$

1.3.2 Root Mean Square (RMS)

The root mean square difference between two data sets A and B is defined as

$$\frac{\sqrt{\sum\limits_{i=1}^{N}(a_i-b_i)^2}}{\sqrt{N}\cdot\max(\max(|a_i|),\max(|b_i|))}$$

1.4 Running Tests

To run all test with different settings

./batch_run.sh -d <dest dir>

which runs all the configs in dest dir with different combinations of $-pv|-pv_with_gpu \pm -x2$ and four random number ranges

- r0: [-1, 1]
- r1: [-10, 10]
- r4: [1, 5]
- r5: [5, 10]

The results are written into verify_f16_\${SET}_\${RAND_RANGE}.txt. For example, file verify_f16_xdlops_pv_rand0.txt contains the results for xdlops gpu kernel with cpu validation using [-1, 1] as the random number range.

To run a single test

./batch_run.sh -d <dest dir> -s <bad test>

which runs the same combinations as in batch mode but only for the single test <dest dir>/<bad test>. The results are written to bad_test_\${BAD_TEST}.txt.

2 Experiment Results

2.1 Effects of Input Number Magnitude

The following table shows the aggregated values among all fp16 tests for all metrics on MI200. The error with r1 ([-10, 10]) is larger than the values with r0 ([-1, 1]) except for the maxRelDiff and RMS metrics. This is because with r0, the denominator of the relative difference is much closer to zero than with r1. When small denominators are ignored, which leads to the maxRelDiff old metric, the error with r1 is larger than that with r0.

	r0	r1
maxEpsilonDiff ave	314.24	1,627.64
$\max Epsilon Diff \max$	2,030.33	23,316.33
maxAbsDiff ave	0.15	3
$\max Abs Diff \max$	5.67	26.67
maxRelDiff ave	4.42	2.79
$\max RelDiff \max$	111.33	105.67
maxRelDiff old ave	0.02	0.1
maxRelDiff old max	0.1	2.2
RMS ave	2.12E-06	1.31E-06
RMS max	1.55E-05	6.57E-06

Observation: The larger the magnitude, the larger the error. (This is true because r0 and r1 are around 0)

2.2 Effects of Underflow and Overflow

The following table shows the aggregated values for r0 ([-1, 1]), r4 ([1, 5]), and r5 ([5, 10]) on MI200. The errors are much larger with r0 than r4/r5 except for the maxAbsDiff and RMS metrics. The maxAbsDiff metric is expected to be larger with r4 and r5 since larger input numbers are used. The RMS metric is relatively stable with different ranges.

To show the effects of overflow, a test is considered to be pass if its maxEpsilonDiff ≤ 1 and the pass rate among all test for each random range is aggregated in the row pass rate. When r0 is used, most tests have large maxEpsilonDiff values. When r4 is used, the max maxEpsilonDiff value among all tests is 1. Therefore, all tests pass, i.e. the pass rate is 100%. However, the pass rate with r5 is not 100% though the max maxEpsilonDiff value among all tests is 1. This is because 3 tests have overflow (value > 65504) with r5.

	r0	r4	r5
maxEpsilonDiff ave	314.24	0.73	0.72
$\max Epsilon Diff \max$	2,030.33	1	1
pass rate	27.73%	100%	97.82%
maxAbsDiff ave	0.15	2.67	4.99
$\max Abs Diff \max$	5.67	32	32
maxRelDiff ave	4.42	5.64E-04	4.69E-04
$\max RelDiff \max$	111.33	9.80E-04	9.80E-04
maxRelDiff old ave	2.06E-02	5.64E-04	4.69E-04
maxRelDiff old max	9.50E-02	9.80E-04	9.80E-04
RMS ave	2.12E-06	5.34E-06	3.21E-06
RMS max	1.55E-05	2.47E-05	1.25E-05

Observation:

- 1. When subnormal numbers are not involved, much smaller errors are expected.
- 2. The RMS metric is insensitive of subnormal numbers.
- 3. Larger inputs can lead to overflow in some tests.

2.3 Effects of validation methods

The following table shows the results for all metrics with different combinations of kernel and validation methods. (The table does not show results for nonxdlops+gpu because the error is always 0 under all metrics.) Note that gpu validation function does not enable xdlops.

The cpu validation function implements the convolution operation in a sequential manner and uses fp64 variable as the accumulator. The nonxdlops gpu kernel converts the convolution to gemm and implements the gemm without using the mfma instructions. The accumulator is fp32? The xdlops gpu kernel converts the convolution to gemm and implements the gemm with mfma instructions. The fundamental hardware is the DOT4_F32_F16 unit inside the miSIMD Matrix Core, which performs a fused multiply and add of 4 pairs of elements and accumulates the result with FP90.

	xdlops+cpu	xdlops+gpu	nonxdlops+cpu
maxEpsilonDiff ave	352.98	818.53	285.98
$\max Epsilon Diff \max$	3,959.25	9,306.00	5,746.25
maxAbsDiff ave	3.13	3.76	1.21
$\max Abs Diff \max$	24.13	28.00	20.13
maxRelDiff ave	2.01	2.82	0.57
$\max RelDiff \max$	68.5	81.50	12.75
maxRelDiff old ave	2.45E-02	4.96E-02	1.72E-02
maxRelDiff old max	3.20E-01	8.43E-01	5.60E-01
RMS ave	2.75E-06	4.57E-06	1.67E-06
RMS max	1.26E-05	2.03E-05	1.16E-05

Observation: xdlops+gpu has the largest error and nonxdlops+cpu has the smallest error.

Explanation:

- 1. CPU implementation and nonxdlops GPU kernel are close to each other because they have similar truncation behavior and computation order.
- 2. CPU implementation and xdlops GPU kernel have medium difference because they have similar truncation behavior but different computation order.
- 3. xdlops and nonxdlops GPU kernels have largest error because they have different truncation behaviors and computation order.

2.4 Difference between MI100 and MI200

The relevant hardware difference between MI100 and MI200 is that the mfma instructions on MI200 flush denormalized numbers to zero in both input and output. (Refer to table 25 of the MI200 Manual² for descriptions of the instructions.) The following table shows the result on MI100 and MI200 for all metrics with different random number ranges.

	MI200(r0)	MI100(r0)	MI200(r1)	MI100(r1)	MI200(r4)	MI100(r4)	MI200(r5)	MI100(r5)
maxEpsilonDiff ave	314.24	90.50	1,627.64	1,757.24	0.73	0.72	0.72	0.72
maxEpsilonDiff max	2,030.33	1,739.00	23,316.33	40,028.00	1.00	1.00	1.00	1.00
maxAbsDiff ave	0.15	0.15	3.00	2.95	2.67	2.67	4.99	4.99
maxAbsDiff max	5.67	5.67	26.67	26.67	32.00	32.00	32.00	32.00
maxRelDiff ave	4.42	1.70	2.79	2.68	5.64E- 04	5.63E- 04	4.69E-04	4.68E-04
maxRelDiff max	111.33	34.00	105.67	92.67	9.80E- 04	9.80E- 04	9.80E- 04	9.80E- 04
maxRelDiff old ave	0.02	0.01	0.10	0.10	5.64E- 04	5.63E- 04	4.69E-04	4.68E-04
$\begin{array}{c} \operatorname{maxRelDiff\ old} \\ \operatorname{max} \end{array}$	0.10	0.09	2.20	2.22	9.80E- 04	9.80E- 04	9.80E- 04	9.80E- 04
RMS ave	2.12E- 06	1.80E-06	1.31E-06	1.31E-06	5.34E- 06	5.33E- 06	3.21E- 06	3.21E- 06
RMS max	1.55E-05	1.55E-05	6.57E- 06	6.77E- 06	2.47E- 05	2.47E- 05	1.25E-05	1.25E-05

Observations:

- 1. The results with r4 and r5 are almost the same on MI100 and MI200.
- 2. With r0, errors on MI100 is smaller than that on MI200.
- 3. With r1, errors on MI100 and MI200 are almost the same except for the maxEpsilonDiff metric, in which the error on MI200 is smaller.

2.5 The "Worst" Test on MI200

We focus on the worst test on MI200, i.e. mlir-miopen-driver/auto-e2e/padding_kernel_gemmK.mlir (CHECK_RESNET50_F16_CONFIG1), which is a forward convolution operation with a total of 205520896 elements in the output.

The following table shows the histogram of relDiff old among all elements in the output with r0. The max relDiff old is 0.055, which is generated from val=0.00101471 and kern=0.001070023.

 $^{^2} https://developer.amd.com/wp-content/resources/CDNA2_Shader_ISA_18 November 2021.pdf$

relDiff_old value interval	number of elements
0	205235692(99.861229%)
(0.1e-6)	0(0.00%)
[1e-6, 1e-5)	0(0.00%)
[1e-5, 1e-4]	0(0.00%)
[1e-4, 1e-3]	244810(0.119117%)
[1e-3, 1e-2)	15725(0.007651%)
[1e-2, 0.1)	925(0.000450%)
[0.1, 1)	0(0.00%)
>= 1	0(0.00%)
max	0.055
val	0.00101471
kern	0.001070023

The following table shows the histogram of epsilonDiff among all elements in the output with r0. The max value is 850, which is generated from val=3.50475E-05 and kern=8.57115E-05.

epsilonDiff value interval	number of elements
0	205235692(99.861229%)
1	259106(0.126073%)
2	11473(0.005582%)
[3, 10]	12039(0.005858%)
[11, 100]	2331(0.001134%)
> 100	255(0.000124%)
max	850
val	3.50475 E-05
kern	8.57115E-05

3 Exploration of a "Better" Validation Function

The following two tables show the effects of some improvements of the validation function on the histograms of the error.

- accumulate 4 refers to the changes that the cpu validation function accumulates 4 multiplication results with fp64 and truncates the sum to fp32 before moving forward.
- flush subnormals refers to the changes that the cpu validation function flushes subnormal numbers ($< 2^{-14}$) in the input to zero.

relDiff old	original	accumulate 4	flush subnormals
0	801676(99.858000%)	802015(99.900226%)	802369(99.944321%)
(0.1e-6)	0(0.000000%)	0(0.000000%)	0(0.000000%)
(1e-6, 1e-5)	0(0.000000%)	0(0.000000%)	0(0.000000%)
[1e-5, 1e-4]	0(0.000000%)	0(0.000000%)	0(0.000000%)
[1e-4, 1e-3)	970(0.120825%)	704(0.087691%)	364(0.045340%)
[1e-3, 1e-2)	60(0.007474%)	60(0.007474%)	0(0.000000%)
[1e-2, 0.1)	$3(0.000374\%)^{'}$	$3(0.000374\%)^{'}$	0(0.000000%)
[0.1, 1)	0(0.000000%)	0(0.000000%)	0(0.000000%)
>= 1	0(0.000000%)	0(0.000000%)	0(0.000000%)
cpuVal == 0	0(0.000000%)	0(0.000000%)	0(0.000000%)
max	0.027	0.028	0.00097
cpuVal	-0.001073837	-0.001074791	0.125244141
gpuVal	-0.001045227	-0.001045227	0.125366211

epsilonDiff	original	accumulate 4	flush subnormals
0	801676(99.858000%)	802015(99.900226%)	802369(99.944321%)
1	1047(0.130416%)	722 (0.089933%)	413(0.051444%)
2	46(0.005730%)	40(0.004982%)	19(0.002367%)
[3, 10]	39(0.004858%)	32(0.003986%)	15(0.001868%)
[11, 100]	8(0.000996%)	7(0.000872%)	0(0.000000%)
> 100	0(0.000000%)	0(0.000000%)	0(0.000000%)
max	$3\dot{5}$	$3\dot{5}$	8
cpuVal	-0.001625061	-0.001625061	9.07E-05
gpuVal	-0.00165844	-0.00165844	9.02E-05

Observations:

- 1. accumulate 4 slightly improves the results under both metrics.
- 2. flush subnormals greatly improves the results under both metrics.

4 Summary

From the experiment results, we can see that the errors for fp16 tests are caused by three sources:

- 1. Subnormal numbers flushed to zero
- 2. Computation order difference due to partition of gemm onto the grid
- 3. Truncation behavior difference during accumulation of intermediate results

The first cause can be solved by avoiding subnormal numbers in the computation. A good candidate is to use [1, 5] or [-5, -1] as the random number range to initialize the inputs. Another candidate is to use the RMS metric, which filters out differences between small numbers.

The computational order for each element in the output is determined by the partition mechanism used by the gemm algorithm. It is not fair, if not impossible, for the validation function to take into account the partition mechanism to keep the same computational order as the GPU kernel.

The truncation behavior is another thing that the validation function cannot emulate. To make things worse, the influence of the truncation behavior to the final result is deeply coupled with the computational order of the elements.

Fortunately, the last two causes can be filtered out with element wise metric with a random number range away from 0 or the RMS metric, which is naturally more robust than element wise metrics. Since the element wise metrics and the RMS metric have their pros and cons, the best verification function should apply both metrics and output flexible diagnostic messages for fp16 tests.

All the experiment results can be found at fp16 experiment results³.

5 Proposal of A New E2E Testing Mechanism

5.1 The Current Mechanism

The current E2E tests are in /mlir/test/mlir-miopen-driver/ and follow a workflow as:

config o miopen-gen o mlir-miopen-driver o mlir-rocm-runner o FileCheck

- config: basic configuration of the convolution operations, which includes
 - direction: fwd, bwd_data, and bwd_weight
 - dimension of input, output, and filter
 - layout of input, output, and filter
 - number of channels and batches
 - paddings, dilations, and strides

 $^{^3} https://amdcloud-my.sharepoint.com/:x:/g/personal/lixzhang_amd_com/EaERwVsy40RGksM0xS-RgKUBkw2wSfXNgrHIK8retNke3A$

- miopen-gen: generation of the convolution kernel and host harness code. It can also generate validation and verification functions as requested. The following information are further added to the config to create a complete test:
 - -t: data type of all elements
 - -rand: pattern of random numbers used for initialization
 - * -rand fixed: a pattern of three hard-coded numbers is used to initialize the inputs.
 - * -rand N: generate random numbers using std::rand() within range [-1, 1] for floats and [-5, 5] for integers
 - · N > 0: random numbers are generated using srand(N)
 - · N = 0: random numbers are generated using srand(time(0))
 - * -rand_type: data type of the random numbers
 - -x2: enable xdlops in the GPU kernel
 - pv or -pv_with_gpu: validation function. If either is specified, a verification function is generated to compare the results between the GPU kernel and the validation function. The metric used is the maxRelDiff_{old} and the thresholds are hard-coded as
 - * 0.25 for fp16 data type
 - * 0.000001 (1e-6) for other data type
- mlir-miopen-driver + mlir-rocm-runner: JIT execute the generated code
- FileCheck: checks the output of the verification function: [0] is fail and [1] is pass

The current CI setup for E2E tests is as follows:

- In PR CI, tests in e2e_for_pr/ are enabled as fixed tests (-rand fixed) with GPU validation (-pv_with_gpu)
- In nightly CI, both fixed and random E2E tests are enabled
 - tests in auto_e2e/ are enabled as random tests (-rand 1) with CPU validation (-pv)
 - tests in auto_e2e/ and misc_e2e/ are enabled as fixed tests (-rand fixed) with GPU validation (-pv_with_gpu)

5.2 The Drawbacks of the Current Mechanism

- 1. The old validation/verification method
 - The element wise metric maxRelDiff_{old} is too sensitive to the absolute value of the element.
 - The range of random numbers is hard-coded
 - The hard-coded thresholds make it inflexible to use different threshold for different tests.
 - The verification function is coupled with the data type of the inputs.
- 2. The CI testing mechanism
 - The inputs of fixed tests (-rand fixed) are too "good" to catch any bugs
 - The PR CI contains too few E2E tests.
 - Many E2E tests are organized according to when they were reported (MLIR tickets). This may lead to repetition of testing and makes it confusing to add more tests.

5.3 Proposals

- 1. Use RMS and element wise metrics together in the verification function.
 - The verification function computes [RMS, maxAbsDiff, maxRelDiff] between gpu kernel and the validation function.
 - The verification function compares the metrics with their corresponding thresholds and outputs 0 or 1 for each metric indicating whether the metric is below its threshold. For example, output of [1, 1, 0] indicates that the RMS and maxAbsDiff value are below their threshold, but maxRelDiff is above the threshold.
 - The verification function can also output more detailed information about the difference, such as the histograms of element wise metrics, elements values corresponding to the max error. In this way, invoking miopen-gen -pv|pv_with_gpu can provide more detailed information for debugging purposes. And such information is ignored by lit when running all tests.
- 2. Move the decision of how to do verification out of miopen-gen so that test writers can determine how to verify the results, such as random number patterns and ranges, verification metrics, and thresholds. This also enables the verification function to focus on comparing the results without worrying about

numerical errors from different data types.

- Make random number range and metric thresholds command line options for miopen-gen.
- Each test (CHECK command) can determine which metric(s) to check and the corresponding threshold(s) for the metric(s). For example, if a developer wants to test a config with fp16 data type and use the RMS metric with a threshold of 1e-5 and ignore the element wise metrics, the test can be written as

```
// RUN: miopen-gen %pv %random_data %xdlops --rand_type float -RMS 1e-5
   -fil layout=kyxc -in layout=nhwc -out layout=nhwk -p -t f16 | mlir-miopen-driver -c |
  mlir-rocm-runner --shared-libs=%linalg_test_lib_dir/libmlir_rocm_runtime%shlibext,%conv_val
// CHECK: [1 {{.*}} {{.*}}]
For f32 tests, the developer also wants to make sure the maxRelDiff metric is below 1e-3, the test
can be written as
// RUN: miopen-gen %pv %random_data %xdlops --rand_type float -RMS 1e-5 -maxRelDiff 1e-3
   -fil_layout=kyxc -in_layout=nhwc -out_layout=nhwk -p -t f32 | mlir-miopen-driver -c |
   mlir-rocm-runner --shared-libs=%linalg_test_lib_dir/libmlir_rocm_runtime%shlibext,%conv_val
// CHECK: [1 {{.*}} 1]
Another way to ignore the metric is to set its threshold to a very high value. And this can be
done by cmake during the building. For example
// RUN: miopen-gen %pv %random_data %xdlops --rand_type float -RMS 1e-5 -maxRelDiff 1e-3 -maxA
   -fil_layout=kyxc -in_layout=nhwc -out_layout=nhwk -p -t f32 | mlir-miopen-driver -c |
   mlir-rocm-runner --shared-libs=%linalg_test_lib_dir/libmlir_rocm_runtime%shlibext,%conv_val
// CHECK: [1 1 1]
And the threshold for maxAbsDiff can be set in mlir/test/CMakeLists.txt as
if (MLIR_MIOPEN_DRIVER_IGNORE_MAXABSDIFF)
  set (MAXABSDIFF THRESOLD 1000)
and the variable is set in /mlir/test/lit.site.cfg.py.in as
config.abs_threshold = @MAXABSDIFF_THRESOLD@
```

- 3. Reorganize the current E2E tests
 - Put all E2E tests in the same directory /mlir/test/mlir-miopen-driver/e2e/. We can further group E2E tests based on the directions, data types, or even which model they come from.
 - Since E2E tests are slow to run and we want the PR CI to be as fast as possible, we could remove E2E tests from PR CI and check all E2E tests in the nightly CI.
 - We can still keep the fixed E2E test stage (or should we remove it?) and use a tighter thresholds for all metrics, since -rand fixed + -pv_with_gpu is expected to work for all data types.
 - For random E2E test stage, we can setup for different data types:
 - For integers and f32, use a random range around 0 and turn on both RMS and element wise metrics.
 - For fp16, use a random range away from 0 (bounce between [-3,-1] and [1,3]) and turn on both RMS and element wise metrics.