Towards a better E2E testing framework

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The Problem and History

MLIR#408: FP16 xdlops tolerance too high?

- ▶ The tolerance for fp16 tests is set to 15%
- ► For other datatypes, the tolerance is set to 0.0001%

PR#694: Changed verification threshold for fp16 to 0.0001%

- ► All fp16 tests passed in PR CI
- ightharpoonup Failed in Nightly CI \Rightarrow failures happened in the Random Tests stage

PR#696: Bump fp16 tolerance to 25%

- ▶ We could not even get back to 15%
- ► However, later it turned out all tests passed with 10%

Why do some fp16 E2E tests fail?



Experiment Setup

Sources of configs: tests from auto_e2e/, e2e_for_pr/, and misc_e2e/

- ► Tests in e2e_for_pr/ are enabled as fixed tests with GPU validation in PR CI
- ► Tests in auto_e2e/ are set as random tests with CPU validation in the Nightly CI
- ▶ Tests in auto_e2e/ and misc_e2e/ are also set as fixed tests with GPU validation in the Nightly CI

Settings of the test

- -rand 1: Generate random numbers as inputs
- -rand_min and -rand_max: choose the range of the random numbers
- -pv and -pv_with_gpu: choose the validation method
- ► -x2: enable xdlops

Hardware

- ► MI100 gfx908 ixt-rack-112
- ► MI200 gfx90a lockhart5



How to measure the difference between two outputs (val vs. kern)

Element wise difference

- Absolute difference maxAbsDiff = $\max_{i}(|val_i kern_i|)$
- ► Relative difference (ignore zero denominators)

$$\mathsf{maxRelDiff} = \mathsf{max}(\frac{|\mathsf{val}_i - \mathsf{kern}_i|}{|\mathsf{val}_i|}), \ |\mathsf{val}_i| > 0$$

Relative difference (ignore small denominators)

$$\mathsf{maxRelDiff}_{\mathsf{old}} = \mathsf{max}_i(\frac{|\mathsf{val}_i - \mathsf{kern}_i|}{|\mathsf{val}_i|}), \ |\mathsf{val}_i| > 1 \times 10^{-3}$$

Epsilon difference

$$\max \mathsf{EpsilonDiff} = \max_{i} \left(\frac{|\mathsf{val}_i - \mathsf{kern}_i|}{\epsilon_{|\mathsf{val}_i|}} \right)$$

where $\epsilon_{|val_i|}$ is the precision of fp16 numbers around val_i.



How to measure the difference between two outputs (val vs. kern)

RMS: the root mean square difference between two data set A and B

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

Experiment Results and Observations

- Four random number ranges are chosen:
 - r0: [-1, 1]
 - r1: [-10, 10]
 - r4: [1, 5]
 - r5: [5, 10]
- Numbers in the following tables are aggregated among all tests for all settings

Experiment Results and Observations — Input number magnitude

| | r0 | r1 |
|--------------------|----------|-----------|
| maxEpsilonDiff ave | 314.24 | 1,627.64 |
| maxEpsilonDiff max | 2,030.33 | 23,316.33 |
| maxAbsDiff ave | 0.15 | 3 |
| maxAbsDiff max | 5.67 | 26.67 |
| maxRelDiff ave | 4.42 | 2.79 |
| maxRelDiff 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 |

The left table lists aggregated results on MI200

- ► In general, errors with r1 ([-10, 10]) is larger than that of r0
- Exceptions are maxRelDiff and RMS because the denominators are smaller with r0

Experiment Results and Observations — Underflow and overflow

| | r0 | r4 | r5 |
|--------------------|----------|----------|----------|
| maxEpsilonDiff ave | 314.24 | 0.73 | 0.72 |
| maxEpsilonDiff max | 2,030.33 | 1 | 1 |
| pass rate | 27.73% | 100% | 97.82% |
| maxAbsDiff ave | 0.15 | 2.67 | 4.99 |
| maxAbsDiff max | 5.67 | 32 | 32 |
| maxRelDiff ave | 4.42 | 5.64E-04 | 4.69E-04 |
| maxRelDiff 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 |

The left table lists aggregated results on MI200

- ▶ In general, errors with r0 ([-1, 1]) is larger than that of r4 ([1, 5]) and r5 ([5, 10])
- Exceptions are maxAbsDiff and RMS because the magnitude of inputs with r4/r5 are lager
- ▶ A test passes if maxEpsilonDiff ≤ 1
- ► All tests with r4 pass since the maximum maxEpsilonDiff is 1
- ▶ However, some tests with r5 fail even the maximum maxEpsilonDiff is $1 \Rightarrow$ overflow happens

Experiment Results and Observations — Validation methods

| | xdlops+cpu | xdlops+gpu | nonxdlops+cpu |
|--------------------|------------|------------|---------------|
| maxEpsilonDiff ave | 352.98 | 818.53 | 285.98 |
| maxEpsilonDiff max | 3,959.25 | 9,306.00 | 5,746.25 |
| maxAbsDiff ave | 3.13 | 3.76 | 1.21 |
| maxAbsDiff max | 24.13 | 28.00 | 20.13 |
| maxRelDiff ave | 2.01 | 2.82 | 0.57 |
| maxReIDiff max | 68.5 | 81.50 | 12.75 |
| maxRelDiff old ave | 2.45E-02 | 4.96E-02 | 1.72E-02 |
| maxReIDiff 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 |

The left table lists aggregated results on MI200

- cpu and non-xdlops gpu are closest to each other
- cpu and xdlops gpu have medium difference
- xdlops gpu and non-xdlops gpu have the largest difference

cpu: sequential implementation of the convolution operation gpu non-xdlops: convert convolution to gemm and compute gemm without mfma gpu: convert convolution to gemm and compute gemm with mfma

Experiment Results and Observations — MI200 vs. MI100

mfma instructions on MI200 flush denormalized numbers to zero

| | 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 |
| maxRelDiff old max | 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 |

- ➤ With r4 and r5, MI200 and MI100 behave the same
- ► With r0, MI100 has smaller errors
- With r1, MI200 and MI100 behave similarly

Experiment Results and Observations — Exploration of a "better" validation function

| 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 |

The left table shows the histogram of maxRelDiff_old of the worst test on MI200

- accumulate 4: cpu truncates the sum of every 4 multiplications from fp64 to fp32 ⇒ slightly improves the results
- ▶ flush subnormals: cpu flushes small numbers ($< 2^{-14}$) to zero \Rightarrow greatly improves the results

Summary

Causes of numerical errors for fp16 tests:

- 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

Solutions or workarounds

- Avoid subnormal numbers by choosing a random range away from 0
- Use RMS as the metric

Proposals

- 1. MLIR#620 Apply both RMS and element wise metrics in the verification function
- 2. MLIR#621 Let each test decide how it is passed
 - choose which metric to use
 - choose the tolerance for each metric
- 3. MLIR#619 Auto generate E2E tests at configure or build time
 - Organize all configs in the same .toml file
 - ► For each config, generate tests for all data types, directions, and layouts.
 - ► The MLIR repository only maintains the .tom1 file and the generation script but does not maintain the generated tests.
 - Stop using fixed tests, all tests use -rand 1
 - Check all tests in the Nightly CI and select a subset for the PR CI. Adjust the metrics and their tolerance based on the selected tests.