Image Processing Ops as first class citizens in MLIR: write once, vectorise everywhere!

Prathamesh Tagore

Speaker: Prathamesh Tagore (VJTI)

Authors: Prathamesh Tagore (VJTI), Hongbin Zhang (ISCAS),
Rishabh Bali (VJTI), Yuchen Li (ISCAS)

prathameshtagore@gmail.com

Challenge

Make Image Processing faster 🚄 to make life easier

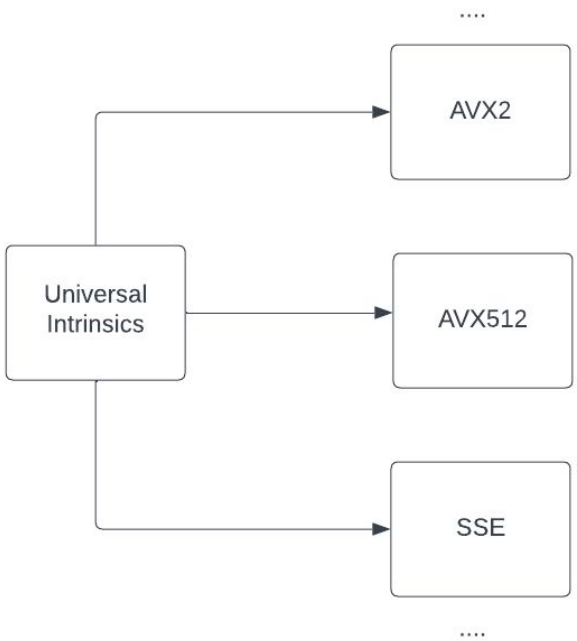
A possible solution

Use Vectorisation?

CPU Vectorisation in Image Processing


```
__m128 dot_product_high = _mm256_extractf128_ps(dot_product, 1);
__m128 dot_product_low = _mm256_castps256_ps128(dot_product);
dot_product_low = _mm_add_ps(dot_product_low, dot_product_high);
dot_product_low = _mm_hadd_ps(dot_product_low, dot_product_low);
dot_product_low = _mm_hadd_ps(dot_product_low, dot_product_low);
```

Universal intrinsics



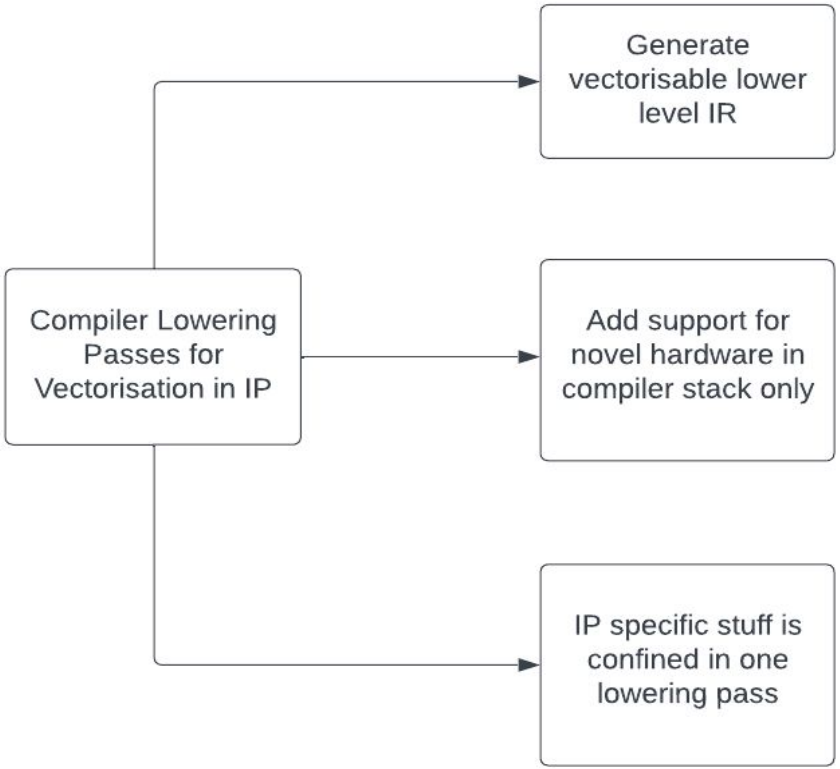
Potential problems while dealing with SIMD intrinsics

- Reliance on a wrapper to provide unified API for targeting each abstraction (Ex. Universal Intrinsics)

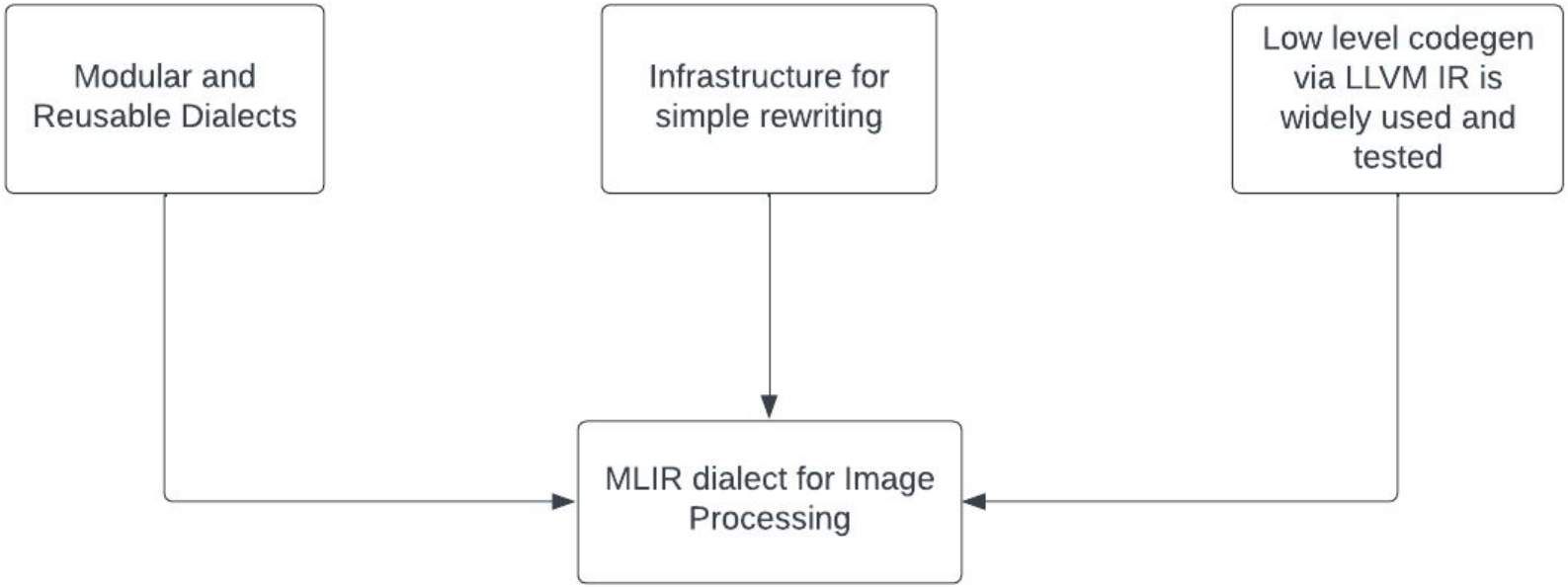
Potential problems while dealing with SIMD intrinsics

- Manual reimplementation of existing algorithms for each hardware technology

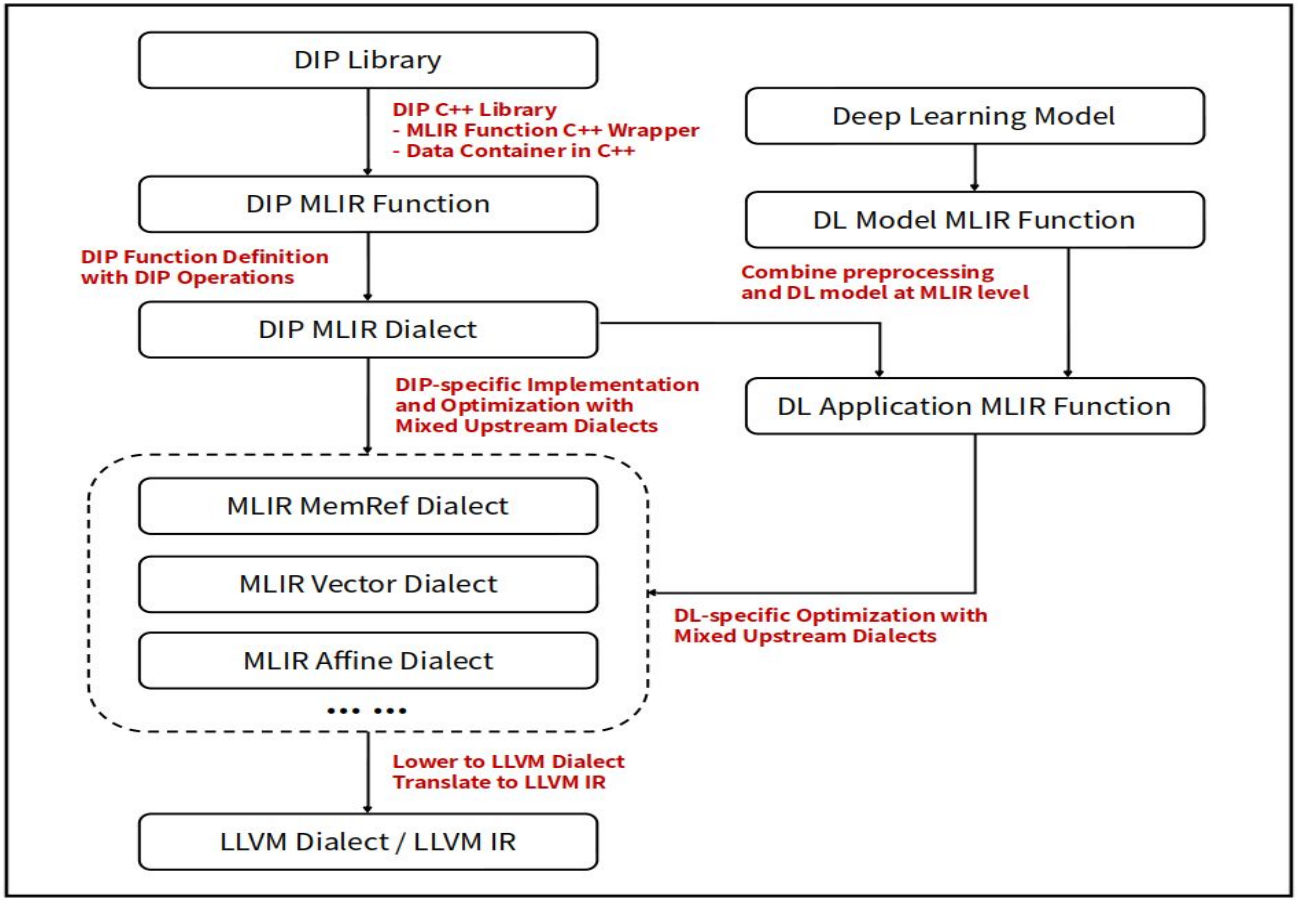
Potential Solution



Why we chose MLIR



DIP (Digital Image Processing) Dialect Overview



DIP Dialect Examples

Original Image

Rotation (45°) Output

Laplacian Filter Output

Resize Output

DIP Dialect Examples

Original Image

Dilation Output

Erosion Output

DL Inference

```
#include <buddy/Container.h>
#include <buddy/DIP.h>
#include <opencv2/opencv.hpp>
… …

// Read image by using OpenCV API.
cv::Mat inputImage = cv::imread("/path/to/dog.png");

// Convert image data into MemRef container.
buddy::Img<float, 4> image(inputImage);

// Define the size of input and output.
intptr_t inputSize[4] = {1, 224, 224, 3};
intptr_t outputSize[2] = {1, 1001};

// Pre-processing for the input image.
buddy::Img<float, 4> input = buddy::resize(image, inputSize);

// Define the output container.
buddy::MemRef<float, 2> output(outputSize);

// The C++ interface wrapped from the MLIR function.
_mlir_ciface_mobilenet_v3(&output, &input);

// Decode the output and print the result.
printResult(output.getData());
```

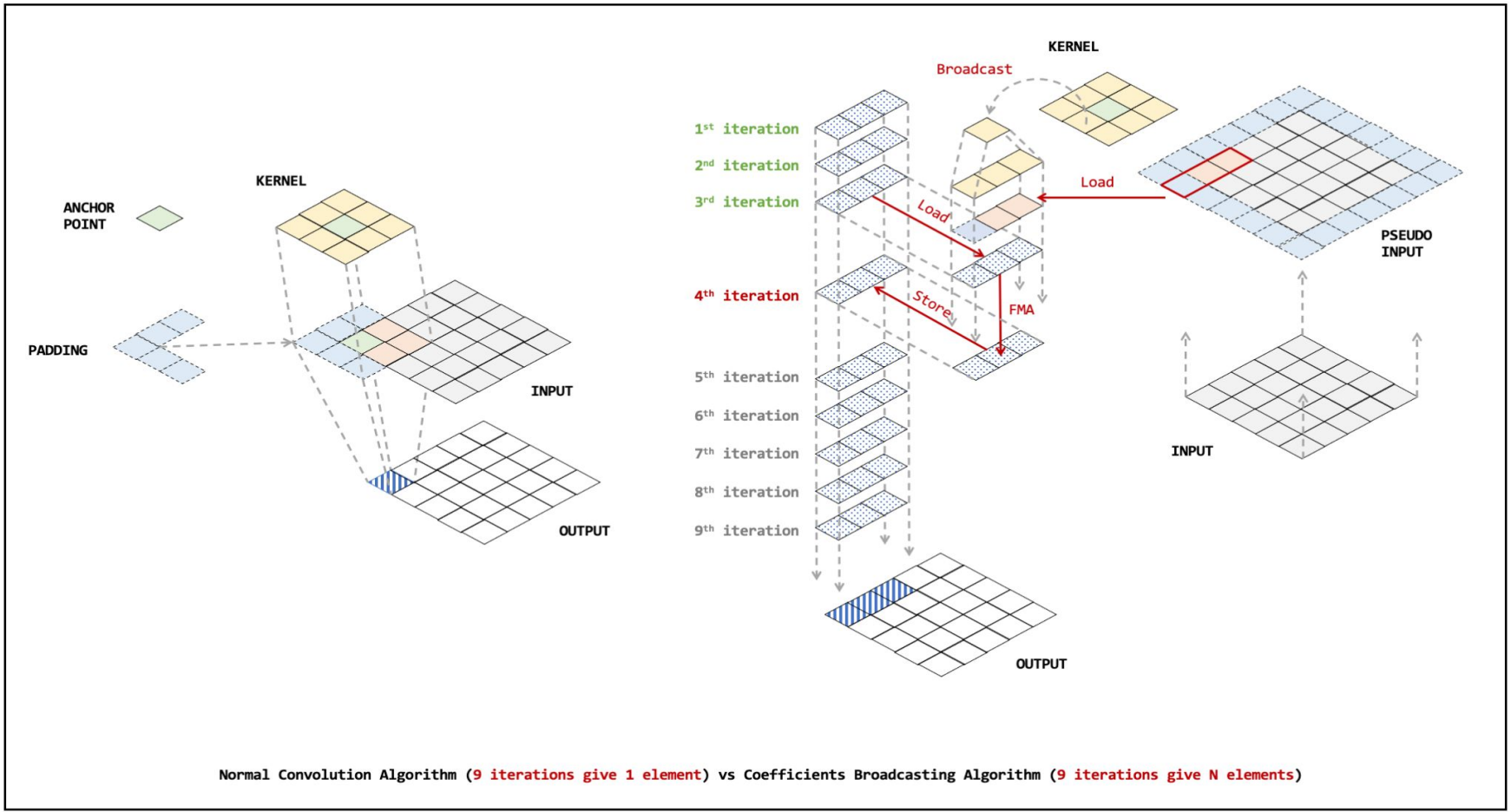
MemRef container in Buddy Compiler is compatible with existing libraries.

The C++ interface is wrapped from custom DIP dialect operations for deep learning pre-processing.

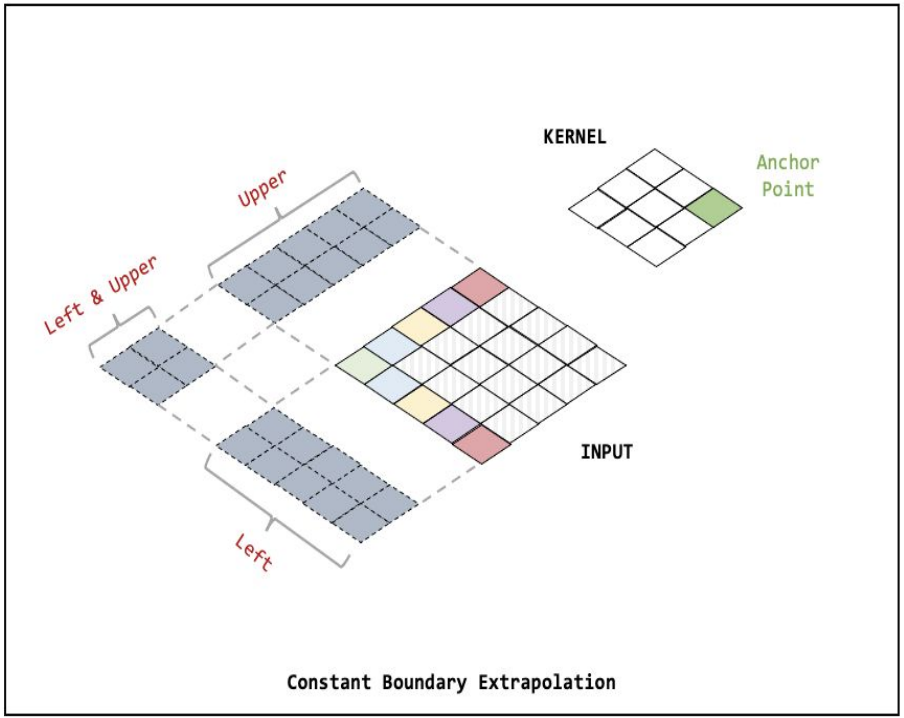
Deep learning model optimized using buddy-mlir.

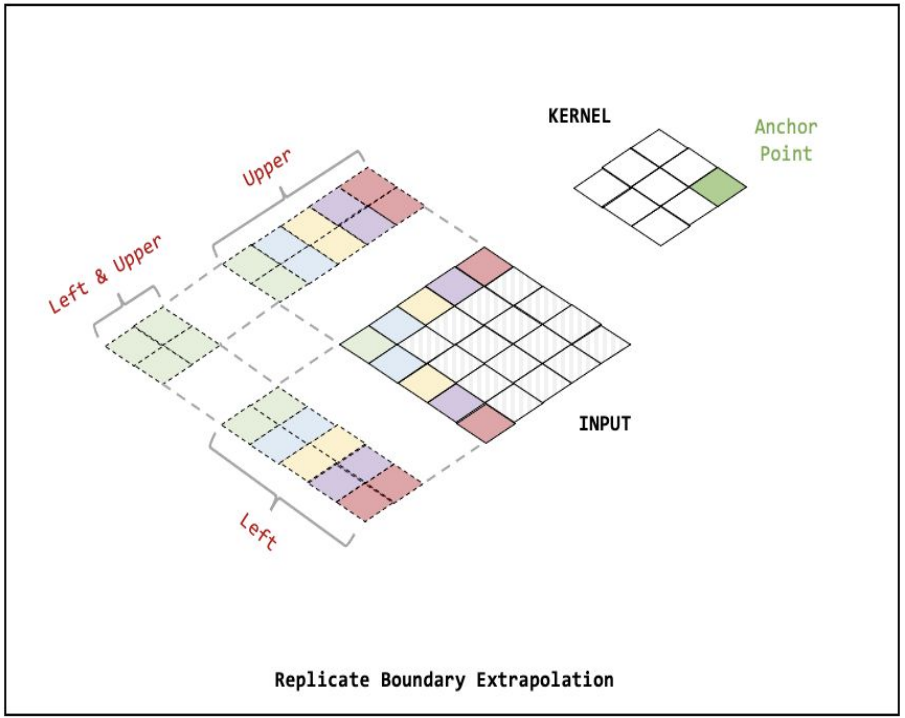
Classification: Samoyed
Probability: 0.529544

Single Channel 2D Image Filtering

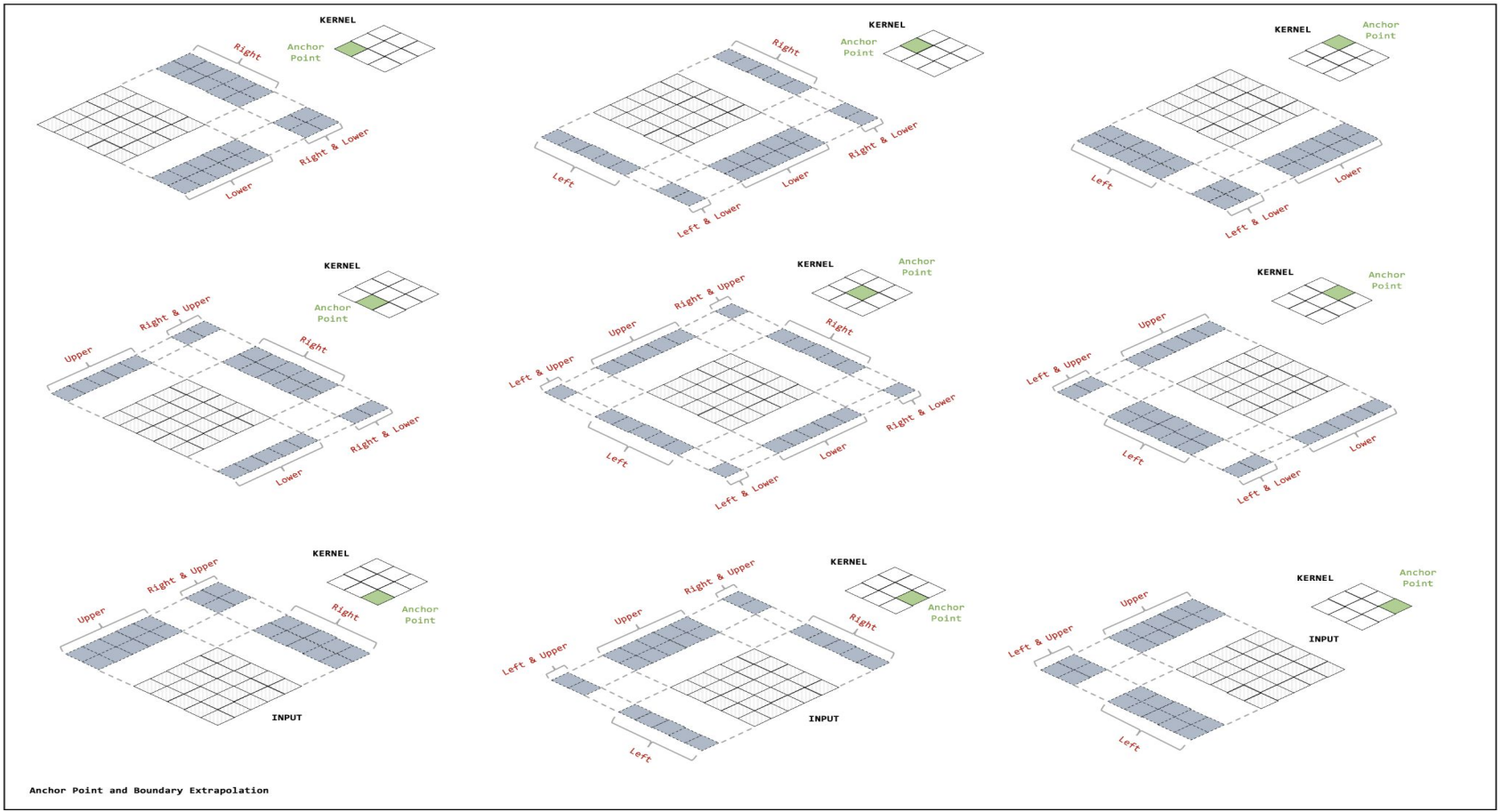


Considered boundary extrapolation strategies for 2D Image Filtering

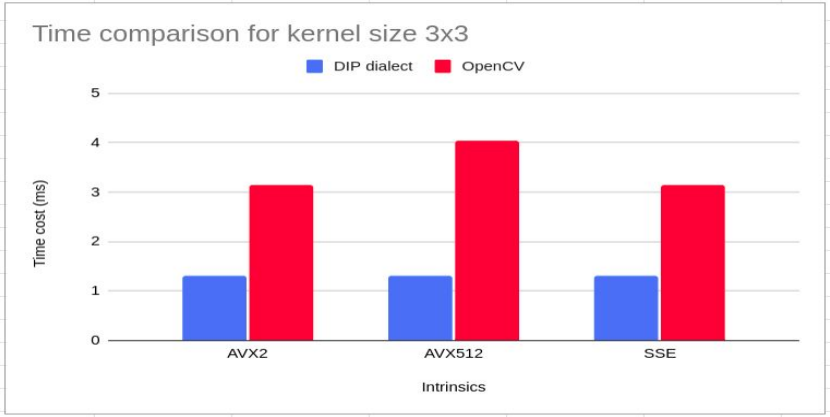


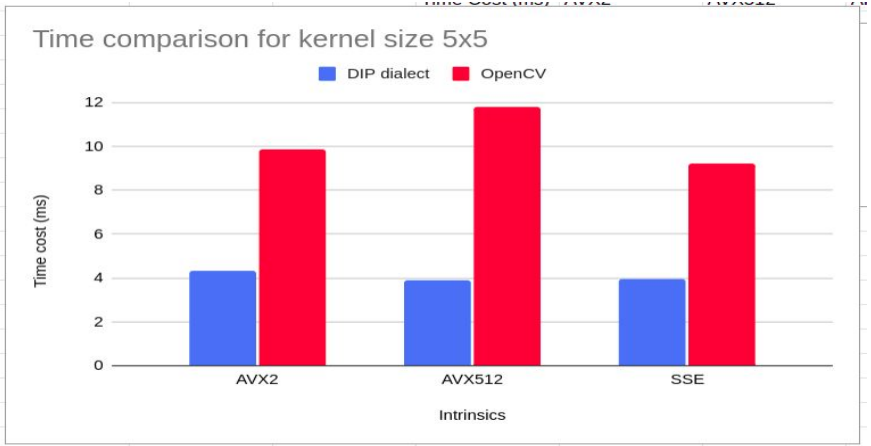


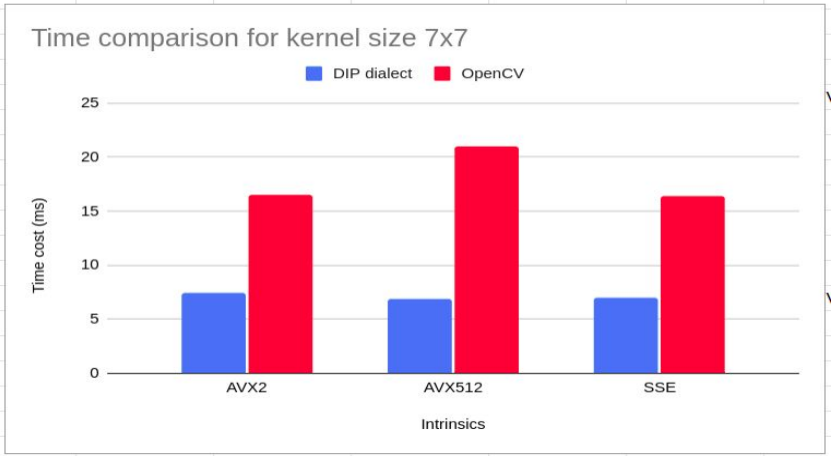
Anchor point position and padding

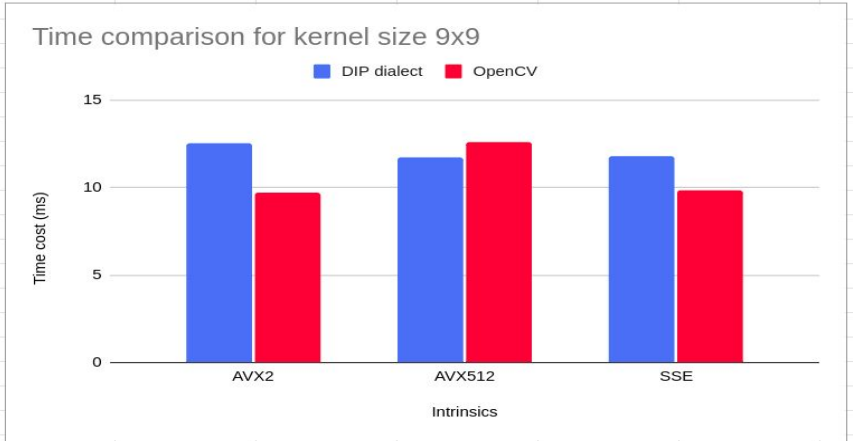


Performance Data

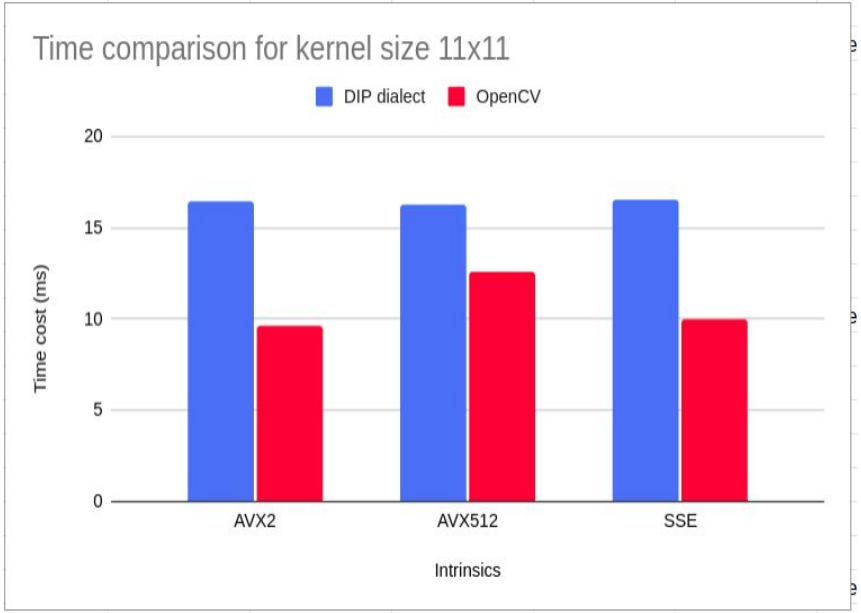


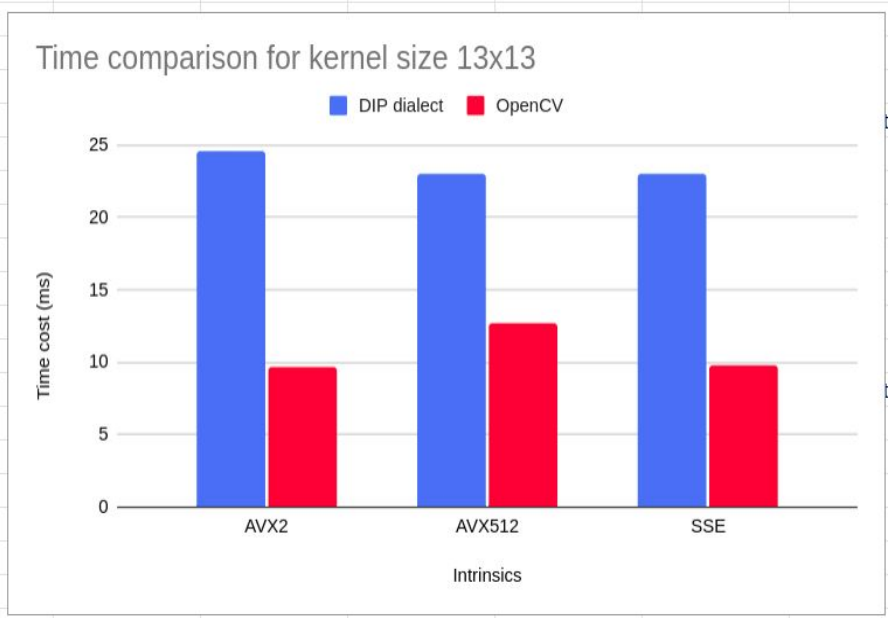




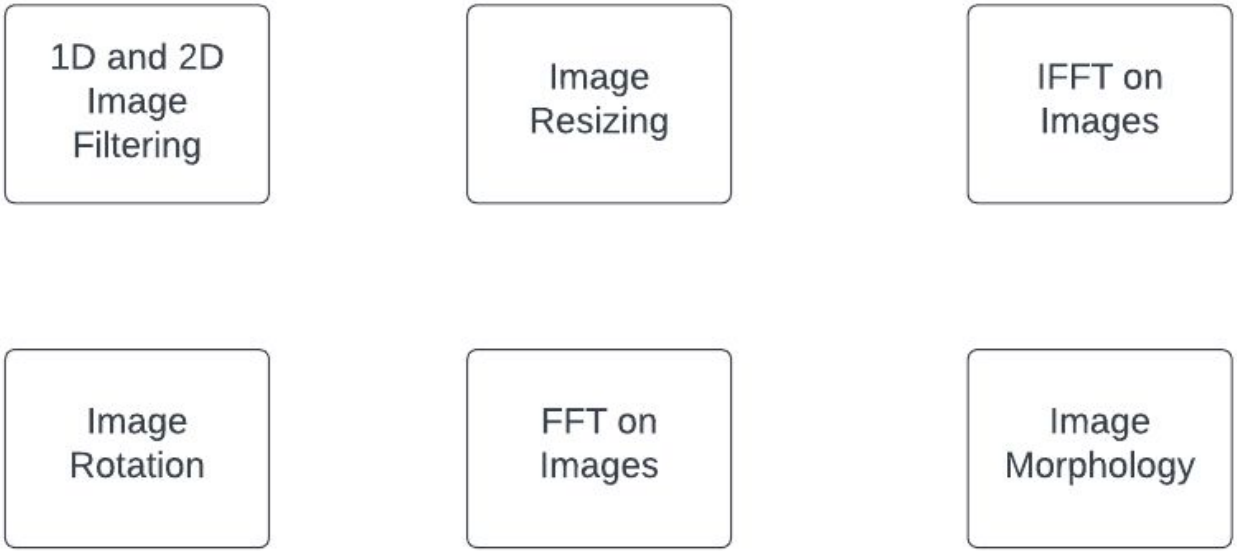


Performance data





Operations currently supported by the DIP dialect:



Potential Scope

Target more hardware?

Integrate with existing MLIR based ML workflows

Interoperability with linalg dialect

Networking

- Email: prathameshtagore@gmail.com
- GitHub: https://github.com/meshtag
- CV: Link
- LinkedIn: https://www.linkedin.com/in/prathamesh-tagore-61aa1a1b1/
- Twitter: https://twitter.com/PrathameshTago1

References

- https://docs.opencv.org/4.x/d6/dd1/tutorial_univ_intrin.html
- https://github.com/buddy-compiler/buddy-mlir
- https://github.com/buddy-compiler/buddy-benchmark
- https://carbon.now.sh/
- Lucidchart