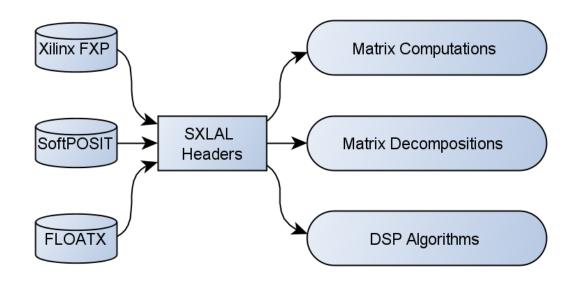
Synthesizable Approximate Linear Algebra Library (SXLAL)

Yun Wu

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Features

- Header-only library without a compiled component
- Various computational precisions support for different arithmetic types
- Synthesizable through Xilinx High Lever Synthesis (HLS)



Supporting Arithmetic

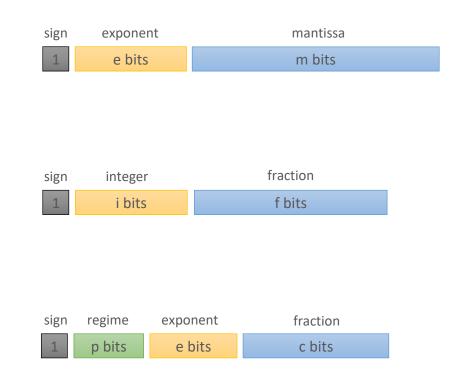
• IEEE 754 floating point arithmetic sign · mantissa · 2^{exponent}

Q format fixed point arithmetic

$$sign \cdot (2^{integer} + 2^{-fraction})$$

Type III Unum – Posit

$$sign \cdot (2^{2^p})^{regime} \cdot 2^{exponent} \cdot \left(1 + \frac{fraction}{2^c}\right)$$



Supporting Linear Algebra

Matrix types:

- Real general (nonsymmetric) real
- Complex general (nonsymmetric) complex
- SPD symmetric positive definite (real)
- HPD Hermitian positive definite (complex)
- SY symmetric (real)
- HE Hermitian (complex)
- BND band

Matrix Operations:

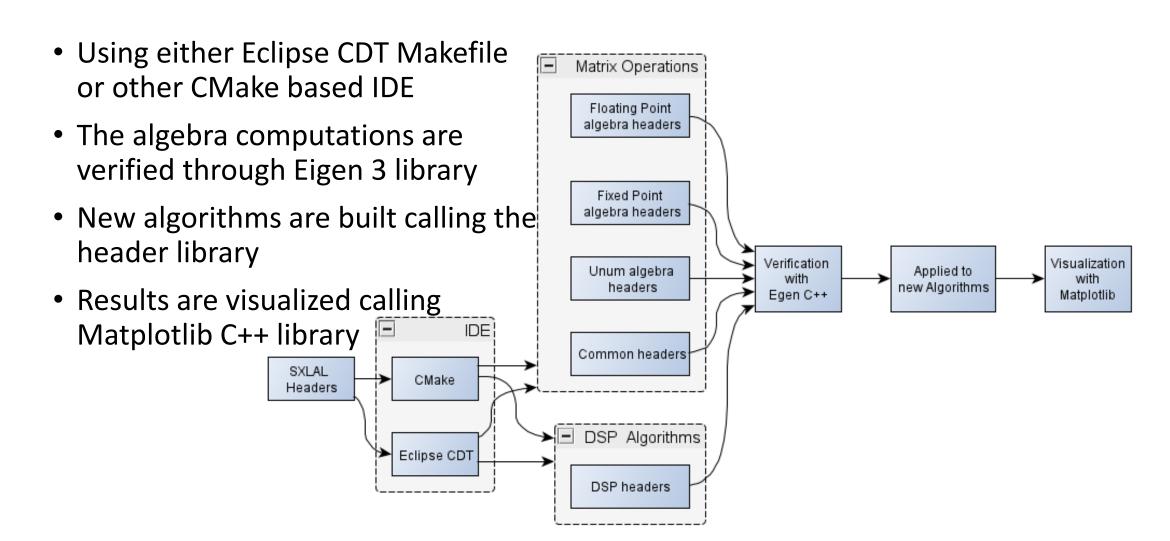
- BA Basic Arithmetic (add, sub, mul, div, inv, etc.)
- TF triangular factorizations (LU, Cholesky)
- OF orthogonal factorizations (QR, QL, generalized factorizations)
- EVP eigenvalue problems
- SVD singular value decomposition
- GEVP generalized EVP
- GSVD generalized SVD

	Real	Complex	SPD	HPD	SY	HE	BND	ВА	TF	OF	EVP	SVD	GEVP	GSVD
Approximate Linear Algebra Library (XLALib)	Yes	No	No	No	No	No	No	Yes	Yes	Yes	No	No	No	No

Prerequisites

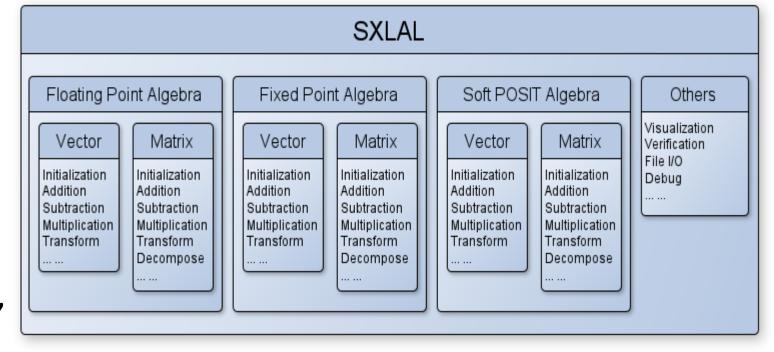
- OS: Linux distribution based on Debian (e.g. Ubuntu 2020.4)
- IDE: Eclipse CDT or CMake
- System libraries: python python-numpy python-matplotlib libboostall-dev libeigen3-dev fftw-dev
- External libraries: FLOATX, Xilinx HLS Fixed Point, SoftPOSIT, Matplotlib C++, Eigen C++

Development & Deployment



Algebra Classes

- Separate header classes for different arithmetic types
- Algebra functions within each header class
- Helper classes
 assisting visualization,
 verification, debug
 and File I/O, etc.



Algebra Functions

• E.g. Matrix Multiplication

arithmetic type

```
float A[4][4], B[4][4], C[4][4];

MAT_MUL <float, 4, 4, 4> (A, B, C);

function name function parameters

MAT_MUL <T, M, N, P> (A, B, C)

MAT_MUL <T, M, N, P> (A, B, C)
```

Function call flow:

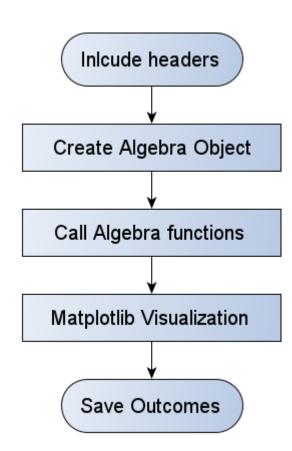
matrix dimensions $C = A \times B$, T A[M][N], T B[N][P], T C[M][P]

Use the algebra function name

Define arithmetic type and data dimension

Library Usage

- Use the header-only library as normal C++ header
- Define specific objects using different arithmetic algebra classes
- Use the algebra functions in the defined object with specific arithmetic type and precision
- Use Matplotlib to visualize the results (Optional)
- Save the results using File I/O classes (Optional)



Example: Proximal Gradient Descent (1)

• The header files of algebra operation using different arithmetic: #include "fpt_algebra.hpp" // Headers for custom floating point linear algebra #include "xfxpt_algebra.hpp" // Headers for fixed point linear algebra #include "softposit_algebra.hpp" // Headers for unum linear algebra

Include header-only library

Define the data structure
 #define ROW 64 // matrix row number
 #define COL 64 // matrix column number
 #define DIAG (COL<ROW ? COL : ROW) // diagonal matrix size

Define data dimension

Create the object to call algebra functions

Float_Point_Algebra Float_Point_Algebra_obj; // Algebra object using floating point arithmetic Fixed_Point_Algebra Fixed_Point_Algebra_obj; // Algebra object using fixed point arithmetic SoftPosit_Algebra SoftPosit_Algebra_obj; // Algebra object using unum arithmetic

Create algebra objects

Example: Proximal Gradient Descent (2)

 Each iteration calling the algebra functions from the header library

```
// Save previous point
Float_Point_Algebra_obj.VEC_EQ<float, DIAG>(x_k_vec, x_k_plus1_vec);
// Compute gradient
float tmp_mul_vec1[DIAG];
Float_Point_Algebra_obj.MAT_VEC_MUL<float, DIAG, DIAG>(
                Amatrix_c, x_k_vec,
                                        tmp_mul_vec1 );
Float Point Algebra obj. VEC_ADD<float, DIAG>( tmp_mul_vec1, bvector_c,
                                                grad g );
// new decent point
float new_point[DIAG];
float tmp mul vec2[DIAG];
Float Point Algebra obj. VEC SCALAR MUL<float, DIAG>( grad g, 1/L c,
                                                         tmp mul vec2 );
Float_Point_Algebra_obj.VEC_SUB<float, DIAG>( x_k_vec, tmp_mul_vec2,
                                                new point );
```

```
algebra
// Proximal projection
float rndnoise[DIAG];
                                                           functions
float a ones vec[DIAG];
float minus_a_ones_vec[DIAG];
Float_Point_Algebra_obj.ONES_VEC<float, DIAG>( a_ones_vec );
Float Point Algebra obj.ONES VEC<float, DIAG>( minus a ones vec );
float tmp_min[DIAG];
float tmp_max[DIAG];
float tmp_mul_vec3[DIAG];
Float_Point_Algebra_obj.RND_VEC<float, DIAG>( rndnoise );
Float_Point_Algebra_obj.VEC_SCALAR_MIN<float, DIAG>(new_point, BOX_CONST, tmp_min);
Float_Point_Algebra_obj.VEC_SCALAR_MAX<float, DIAG>(tmp_min, -BOX_CONST, tmp_max);
Float_Point_Algebra_obj.VEC_SCALAR_MUL<float, DIAG>( rndnoise, error_std,
                                                         tmp_mul_vec3);
Float_Point_Algebra_obj.VEC_ADD<float, DIAG>( tmp_max, tmp_mul_vec3,
                                                x_k_vec);
// check early termination constraint
float norm1, norm2;
float tmp sub vec[DIAG];
Float Point Algebra obj. VEC SUB<float, DIAG>( x k vec, x k plus1 vec,
                                                tmp sub vec);
Float_Point_Algebra_obj.VEC_NORM<float, DIAG>(tmp_sub_vec, norm1);
Float_Point_Algebra_obj.VEC_NORM<float, DIAG>(x_k_vec, norm2);
if( norm1<= EPS_STOP ){</pre>
        break;
```

Calling

Example: Proximal Gradient Descent (3)

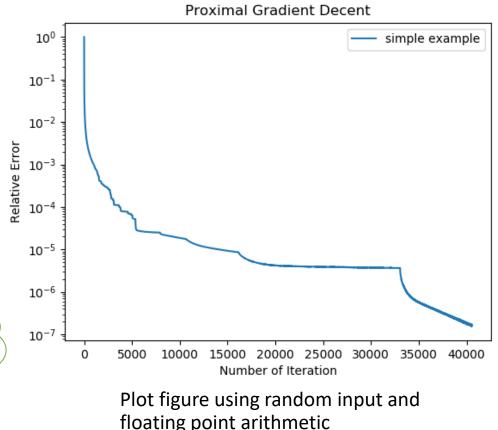
Include matplotlibcpp library

```
#include "matplotlibcpp.hpp"
```

Define matplotlibcpp namespace

```
namespace plt = matplotlibcpp;
```

Plot using matplotlib functions

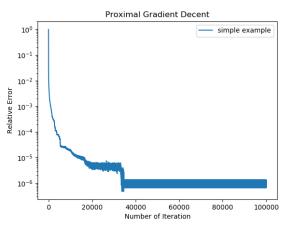


Example: Proximal Gradient Descent (4)

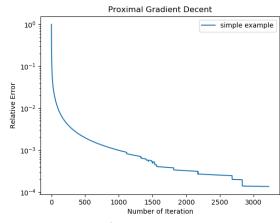
- Custom floating point, e.g.
 using fpx = flx::floatx<5, 10>;
 fpx A[4][4], B[4][4], C[4][4];
 MAT_MUL <fpt, 4, 4, 4> (A, B, C);
- Custom fixed point, e.g.

```
typedef ap_fixed<32,20> fxp;
fxp A[4][4], B[4][4], C[4][4];
MAT_MUL <fxp, 4, 4, 4> (A, B, C);
```

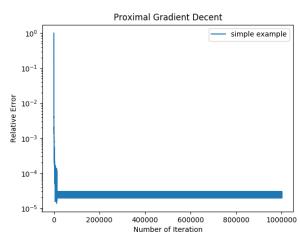
Custom Unum soft POSIT, e.g.
 posit_2_t A[4][4], B[4][4], C[4][4];
 MAT_MUL <posit_2_t, 4, 4, 4, 16 > (A, B, C);



Custom floating point 16 bits



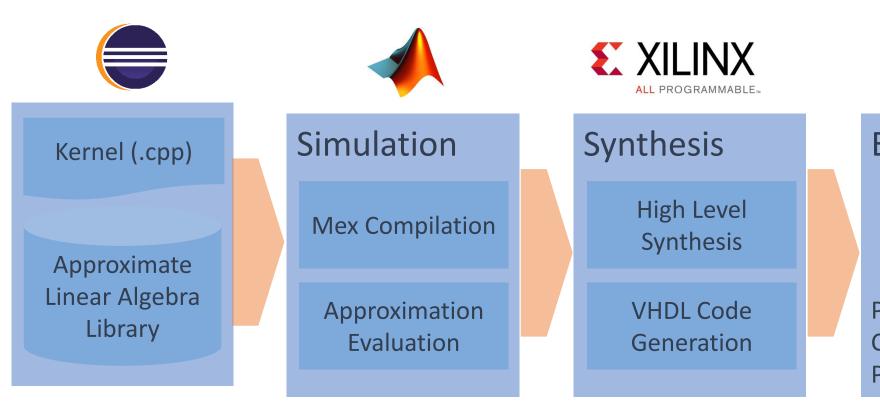
Custom fixed point 28 bits



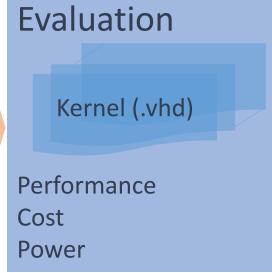
Custom Unum soft POSIT 16 bits

Hardware Prototyping

• C++ kernel \rightarrow MEX verification \rightarrow Xilinx HLS \rightarrow HDL IP verification







Applied Applications

- Convex Optimization Solvers:
 - Proximal Gradient Descent
 - Alternating Direction Method of Multipliers
- Applications:
 - Lidar Depth reconstruction
 - Model Predictive Control (MPC)
 - Proportional–Integral–Derivative (PID) controller

Key Publications

- 1. Efficient Reconfigurable Mixed Precision &1 Solver for Compressive Depth Reconstruction. Journal of Signal Processing Systems 94, 1083–1099 (2022).
- 2. <u>Energy Efficient Approximate 3D Image Reconstruction. IEEE</u>

 <u>Transactions on Emerging Topics in Computing (Early Access)</u>

Future Developing

- EVP eigenvalue problems
- SVD singular value decomposition
- General DSP algorithms
- Complex value linear algebra
- Tensor operations