

Synthesizable Approximate Linear Algebra Library (SXLAL)

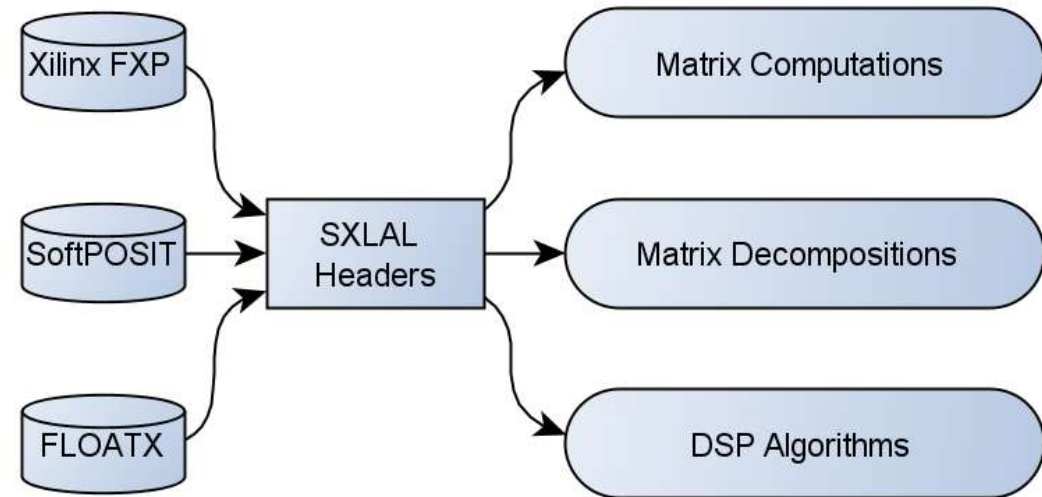
UDRC WP 2.1

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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

$$sign \cdot mantissa \cdot 2^{exponent}$$



- 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 real entries**
- Complex – general complex entries
- 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

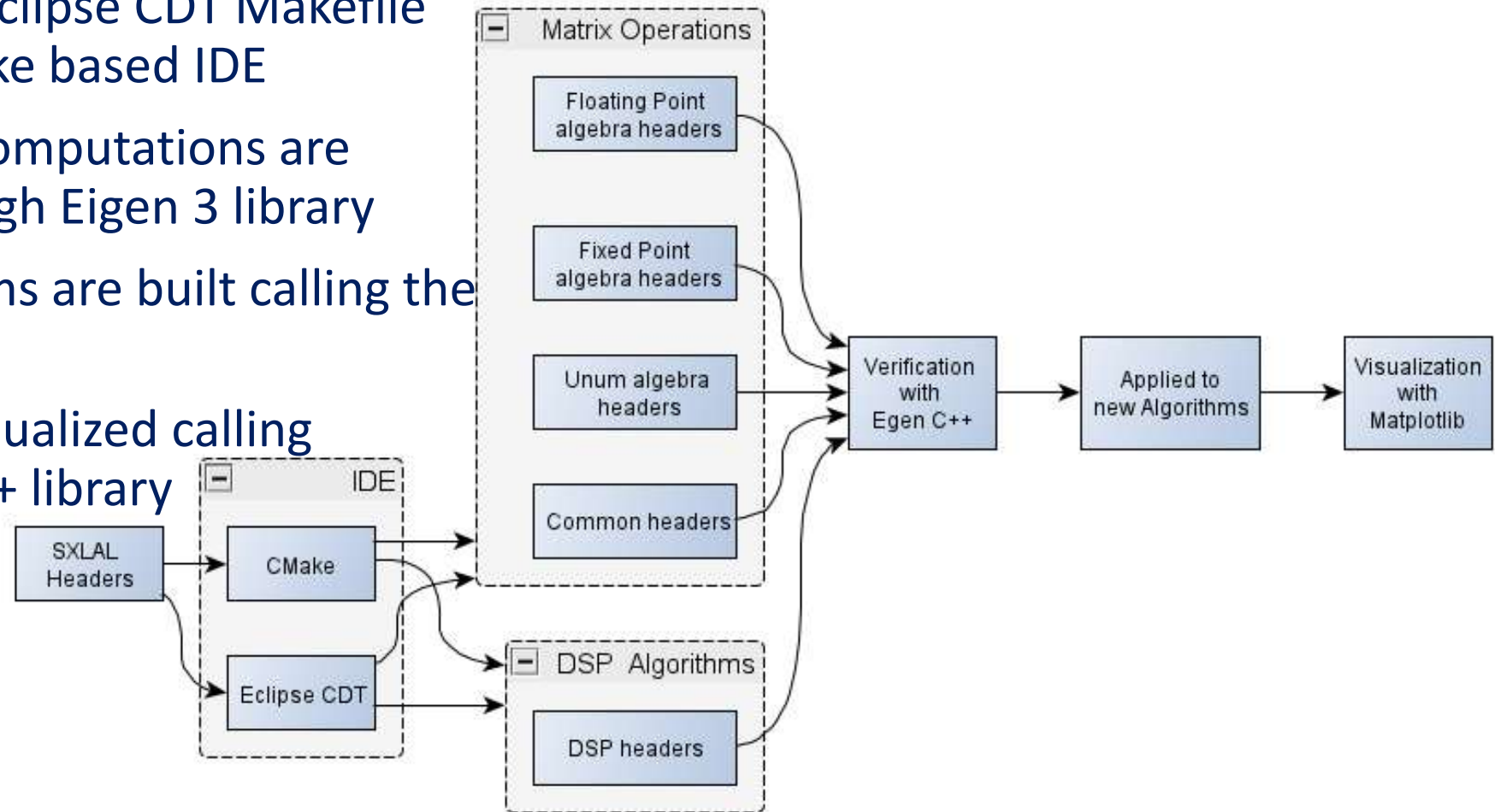
	Real	Complex	SPD	HPD	SY	HE	BND	BA	TF	OF	EVP	SVD	GEVP	GSVD
SXLAL	Yes	No	No	No	No	No	No	Yes	Yes	Yes	No	No	No	No

Prerequisites

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

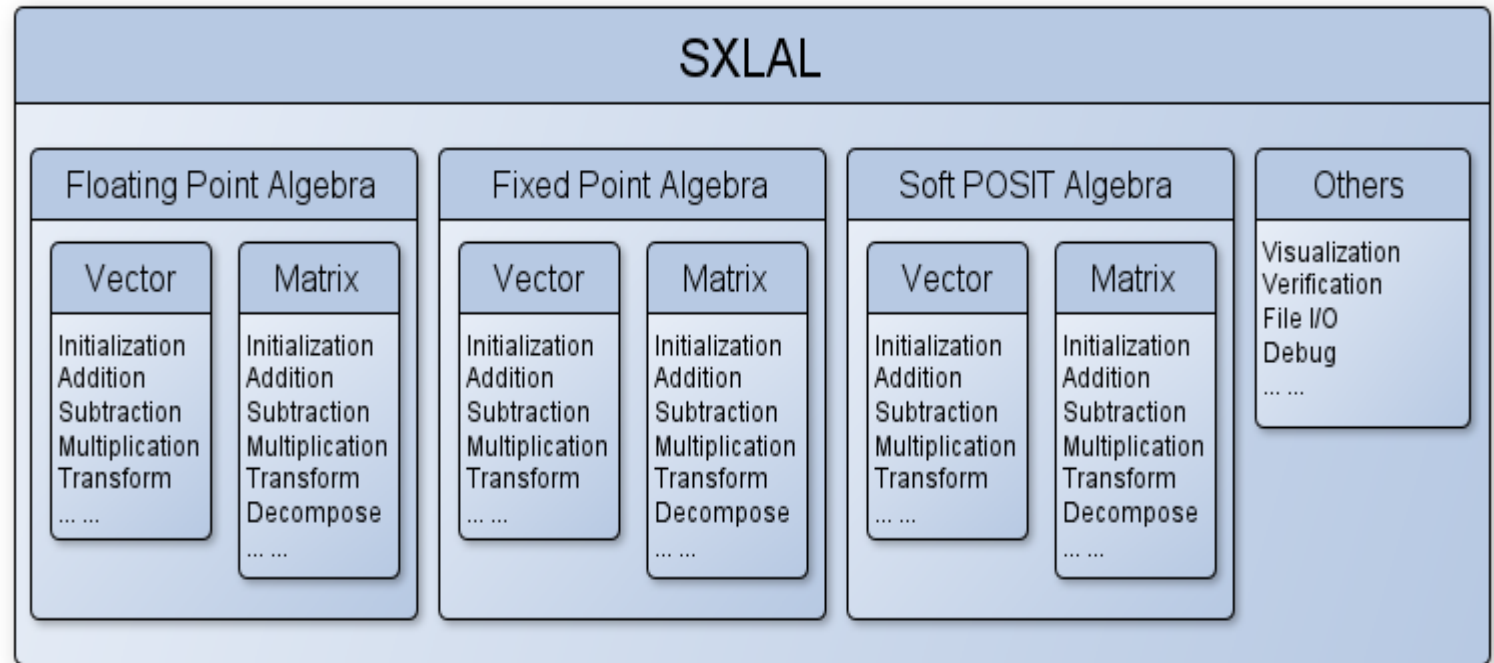
Development & Deployment

- Using either Eclipse CDT Makefile or other CMake based IDE
- The algebra computations are verified through Eigen 3 library
- New algorithms are built calling the header library
- Results are visualized calling Matplotlib C++ library



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.



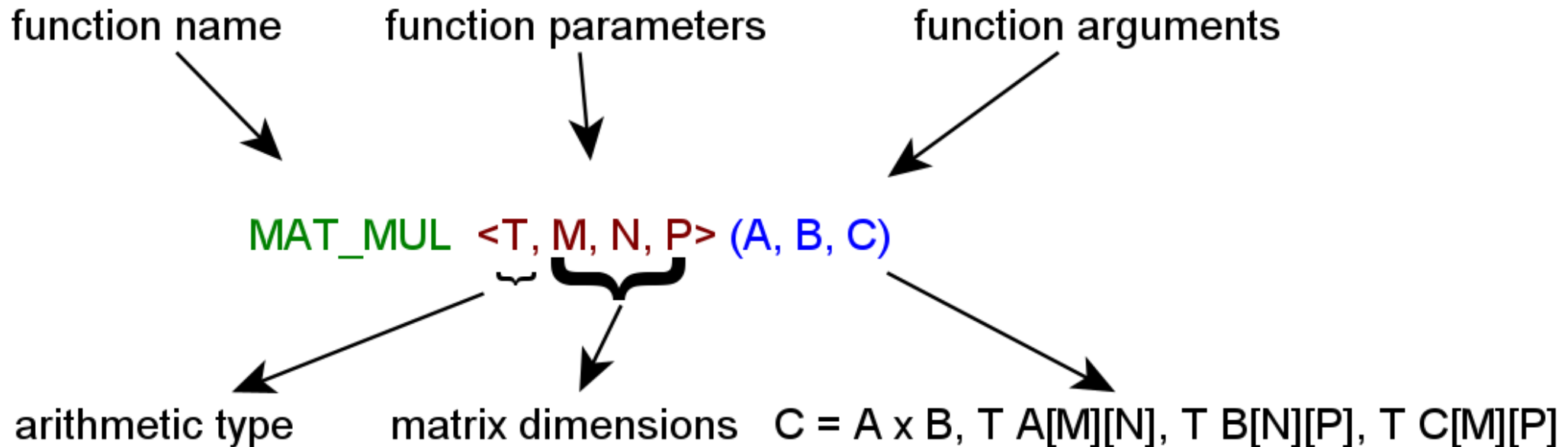
Algebra Functions

- E.g. Matrix Multiplication

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

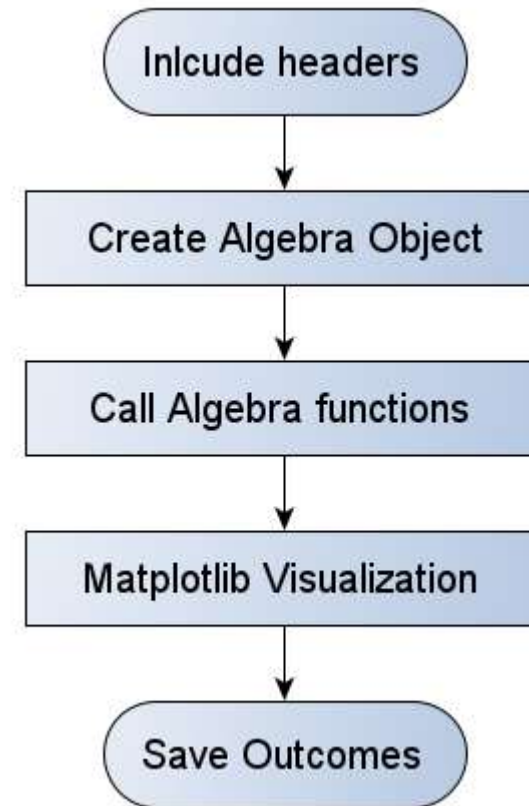
Function call flow:

- Use the algebra function name
- Define arithmetic type and data dimension
- The computation follow the input sequence
- Last argument is the output



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

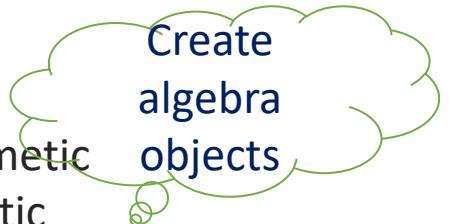
```
#define ROW 64 // matrix row number  
#define COL 64 // matrix column number  
#define DIAG (COL<ROW ? COL : ROW) // diagonal matrix size
```



Define data
dimension

- Create the object to call algebra functions

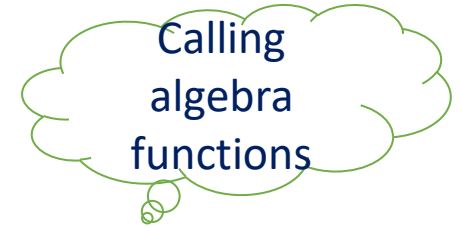
```
Float_Point_Algebra Float_Point_Algebra_obj; // Algebra object using floating point arithmetic  
Fixed_Point_Algebra Fixed_Point_Algebra_obj; // Algebra object using fixed point arithmetic  
SoftPosit_Algebra SoftPosit_Algebra_obj; // Algebra object using unum arithmetic
```



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 );
```

```
// Proximal projection
float rndnoise[DIAG];
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 ){
    break;
}
```

Example: Proximal Gradient Descent (3)

- Include matplotlibcpp library

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

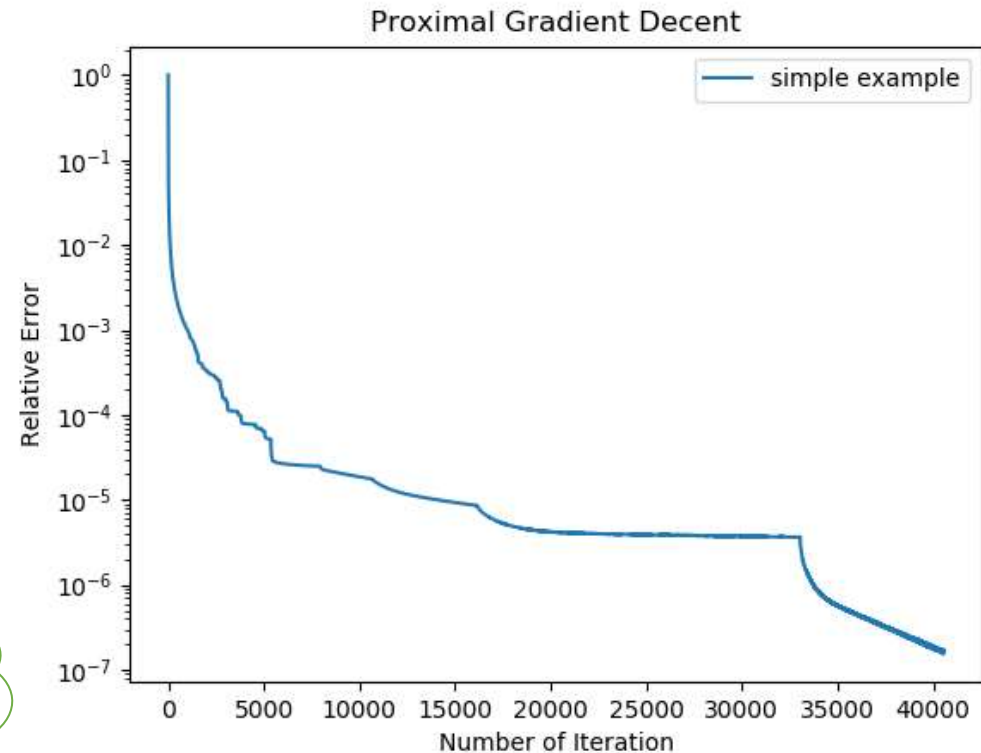
- Define matplotlibcpp namespace

```
namespace plt = matplotlibcpp;
```

- Plot using matplotlib functions

```
#ifdef PLOT_FIGURE  
    // Matplotlib plotting  
    plt::named_semilogy( "simple example", x, y );  
    plt::title("Proximal Gradient Decent");  
    plt::xlabel("Number of Iteration");  
    plt::ylabel("Relative Error");  
    plt::legend();  
    plt::save(figurename);  
#ifdef SHOW_FIGURE  
    plt::show();  
#endif// endif SHOW_FIGURE  
#endif// endif PLOT_FIGURE
```

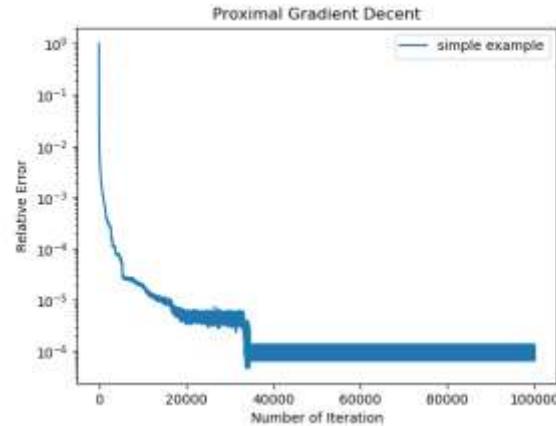
Calling
Matplotlib to
draw figure



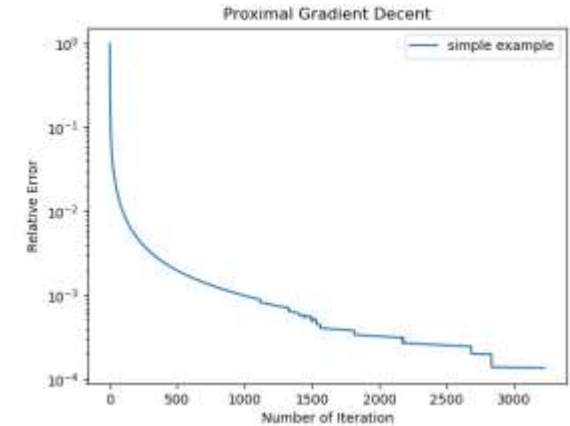
Plot figure using random input and floating point arithmetic

Example: Proximal Gradient Descent (4)

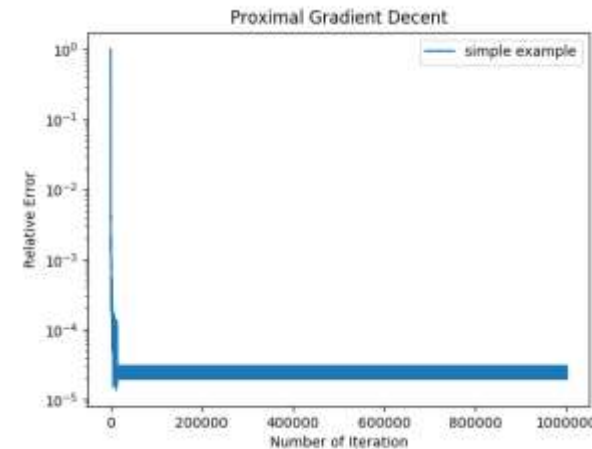
- Custom floating point, e.g.
using `fpx = flx::floatx<5, 10>`;
`fpx A[4][4], B[4][4], C[4][4]`;
`MAT_MUL <fpt, 4, 4, 4> (A, B, C)`;
- Custom fixed point, e.g.
`typedef ap_fixed<32,20> fpx`;
`fpx A[4][4], B[4][4], C[4][4]`;
`MAT_MUL <fxp, 4, 4, 4> (A, B, C)`;
- Custom Unum soft POSIT, e.g.
`posit_2_t 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



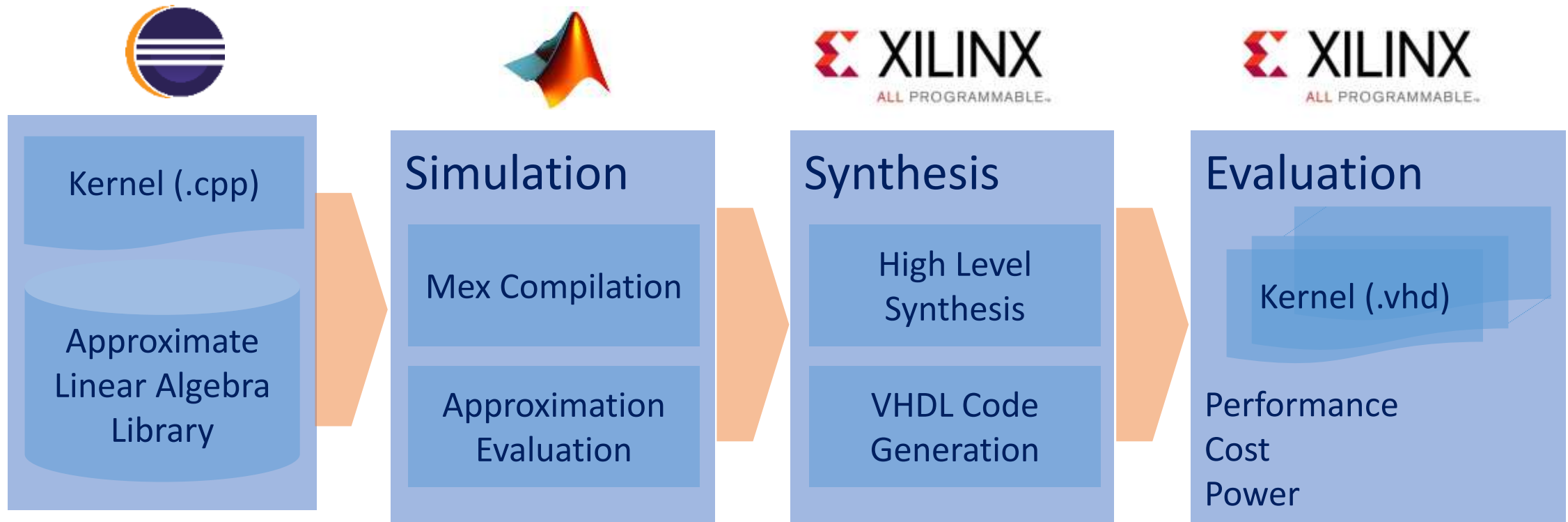
Custom fixed point 28 bits



Custom Unum soft POSIT 16 bits

Hardware Synthesis Flow

- C++ kernel → MEX verification → Xilinx HLS → HDL IP verification



Use Cases

- 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 References

1. Yun Wu, Andrew M. Wallace, João F.C. Mota, Andreas Aßmann, Brian Stewart. [Efficient Reconfigurable Mixed Precision \$\ell_1\$ Solver for Compressive Depth Reconstruction](#). Journal of Signal Processing Systems 94, 1083–1099 (2022).
2. Yun Wu, Andreas Aßmann , Brian Stewart, Andrew M. Wallace. [Energy Efficient Approximate 3D Image Reconstruction](#). IEEE Transactions on Emerging Topics in Computing (Early Access)
3. UDRC WP2.1 Final Report: Resource constrained smart sensor systems. Joao Mota, Andrew Wallace, Yun Wu, Anis Hamadouche.
4. SXLAL GitHub repository: <https://github.com/wincle626/SXLAL>