Michel Steuwer

University of Edinburgh

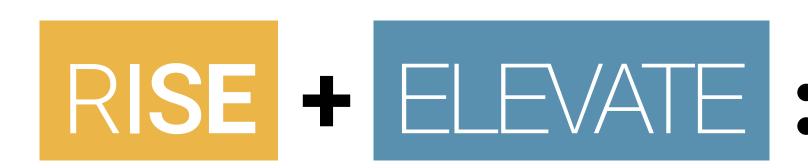


It's all about Choices!

I couldn't agree more with Sven-Bodo about this statement

• I (of course) couldn't disagree more with Sven-Bodo about all the choices made in SaC

Shouldn't the consequence be not to make choices but to offer them?



RISE + ELEVATE: Expose Optimization Choices

RISE A Pattern-Based Intermediate Languages

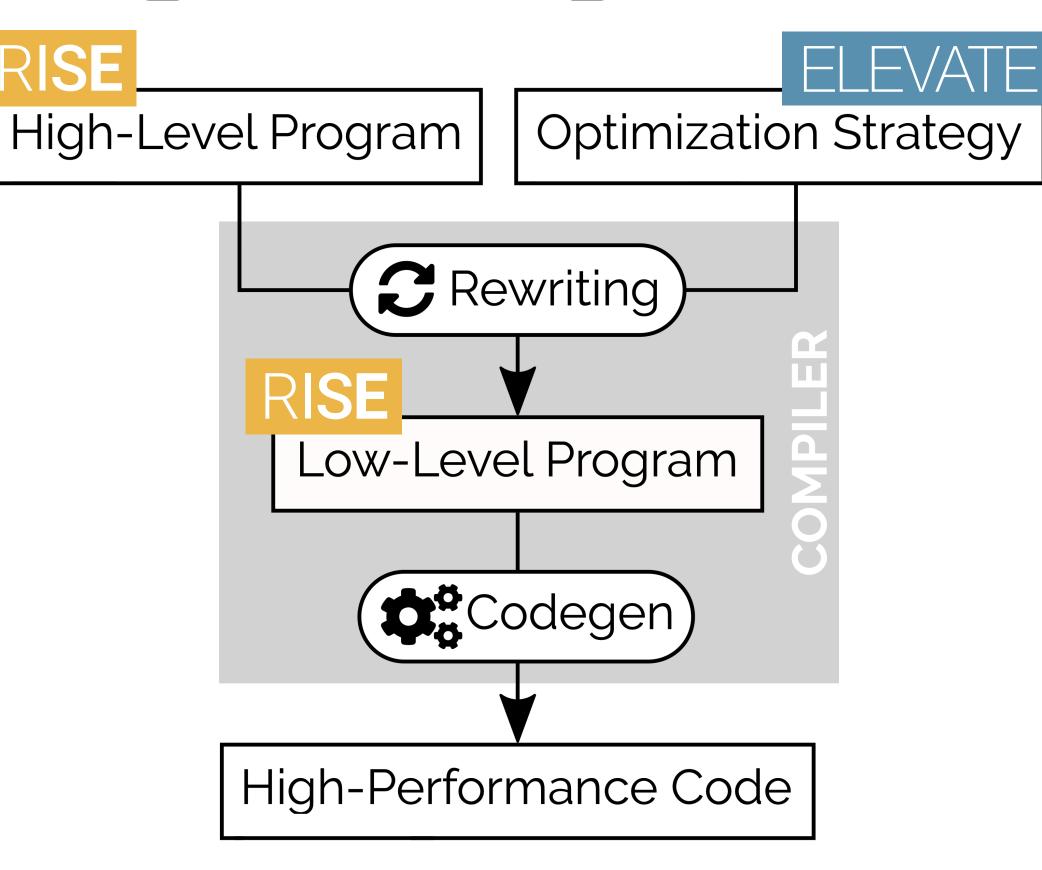
```
Successor
  to LIFT
```

```
def highLevelProgram =
  depFun((n: Nat, m: Nat, o: Nat) ⇒
     fun(A: n.o.f32 \Rightarrow fun(B: m.o.f32 \Rightarrow
       A \triangleright map(fun(row0fA \Rightarrow
          B ▷ map(fun(rowOfB ⇒
            zip(rowOfA)(rowOfB) >
               map(fun(x \Rightarrow fst(x) * snd(x))) \triangleright
                 reduce(add)(0.0f) )) )) ))
```

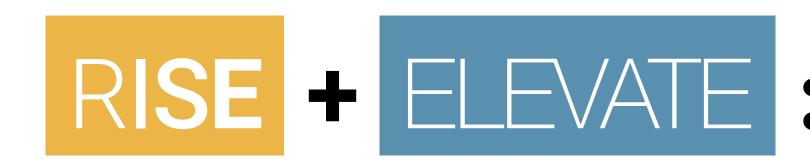
ELEVATE A Programming Language for describing Optimization Strategies

```
def optimizationStrategy =
                     `@` outermost(isMap))
  (`map → mapSeq` `@` outermost(isMap)) `;`
 (`reduce → reduceSeq` `@` everywhere)
```

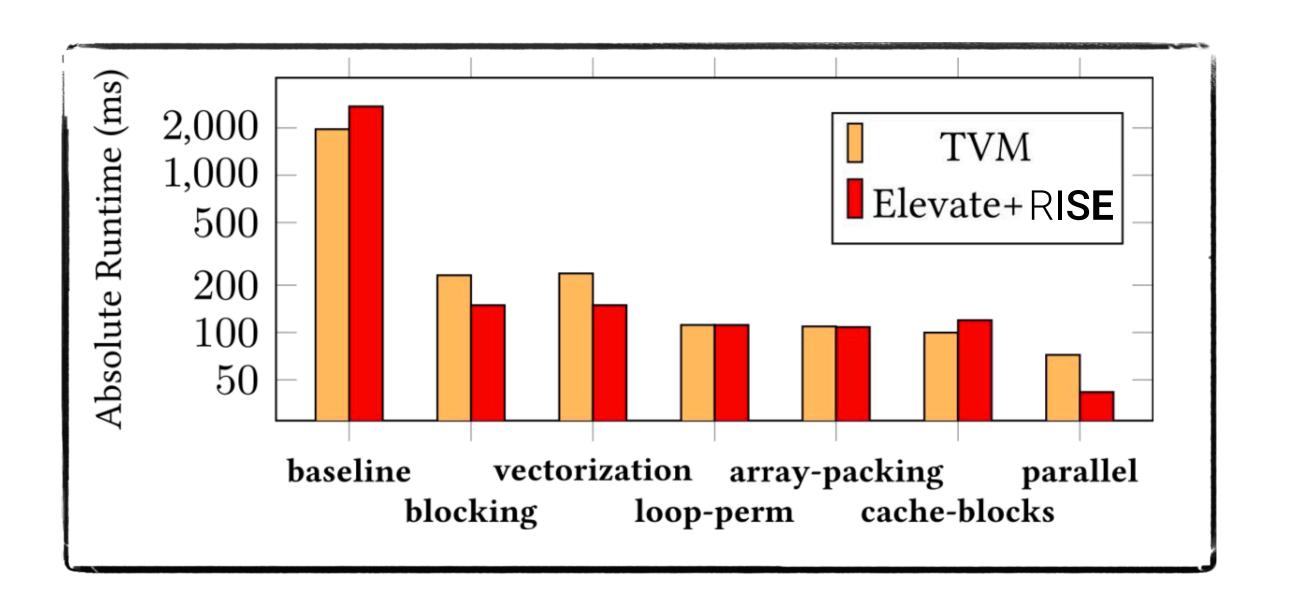
[ICFP 2020] (CGO 2021)

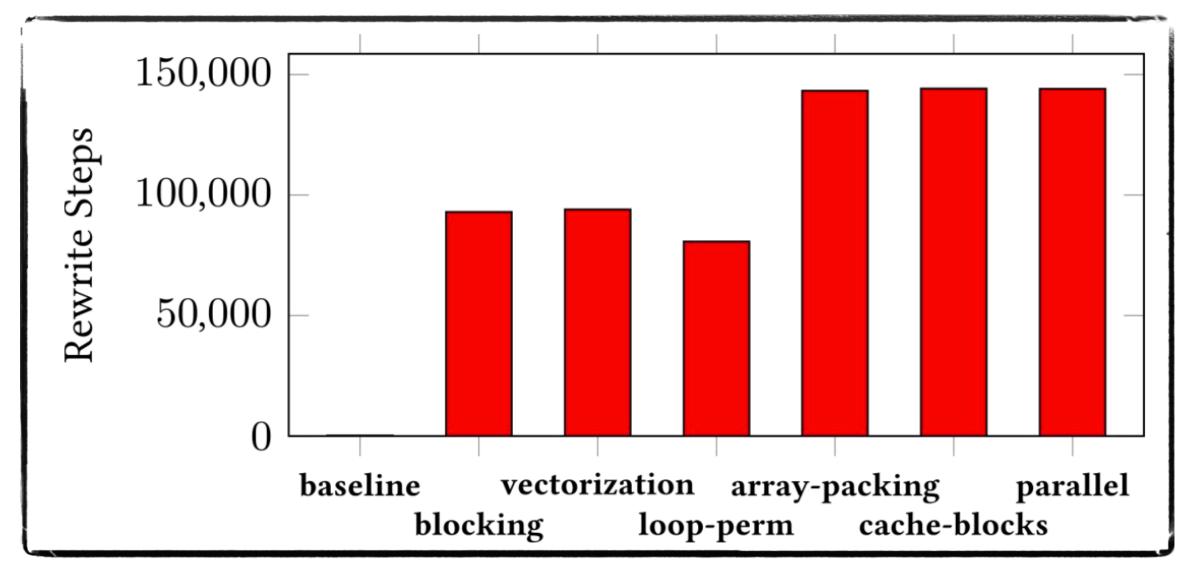


https://rise-lang.org https://elevate-lang.org



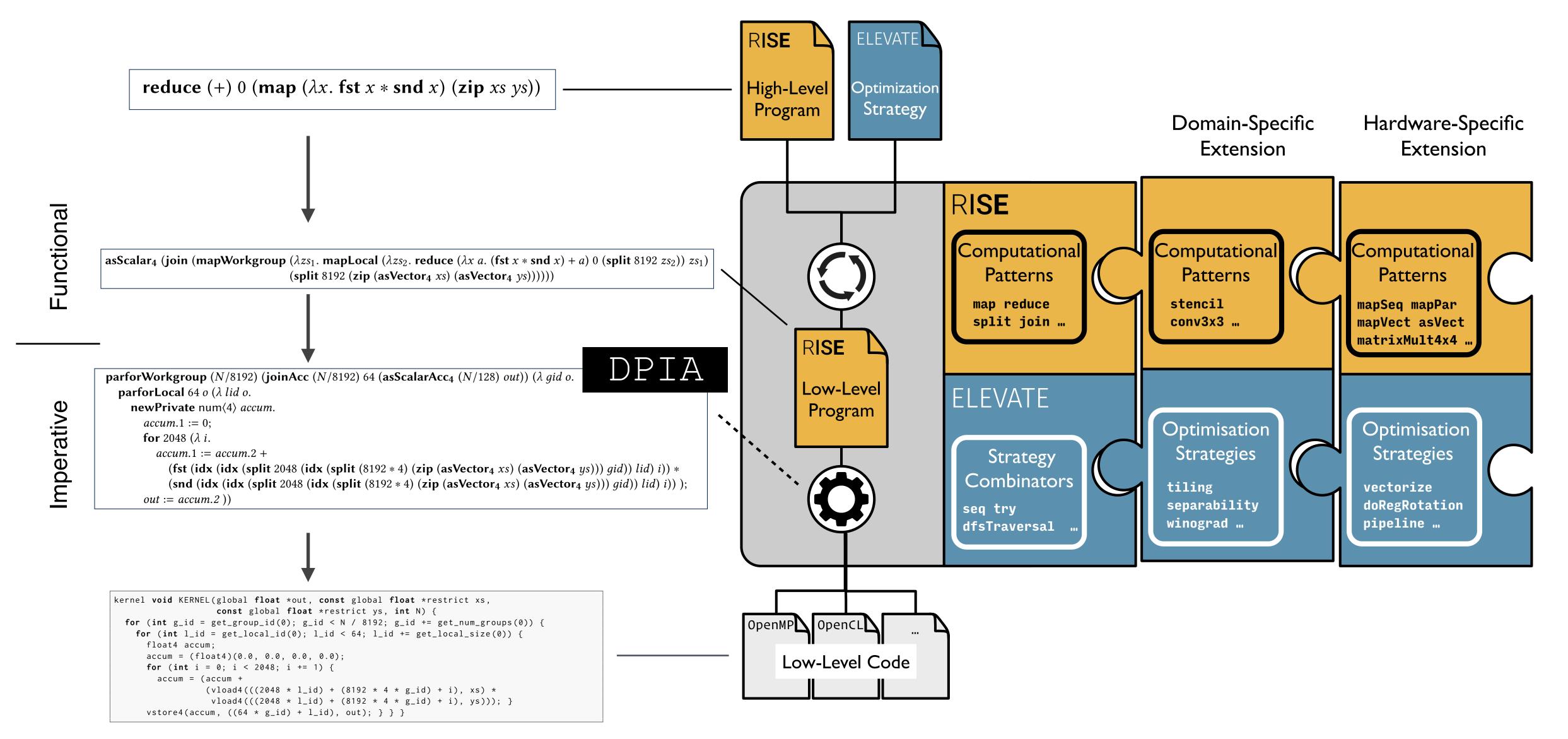
RISE + ELEVATE: Results for Matrix Multiplication







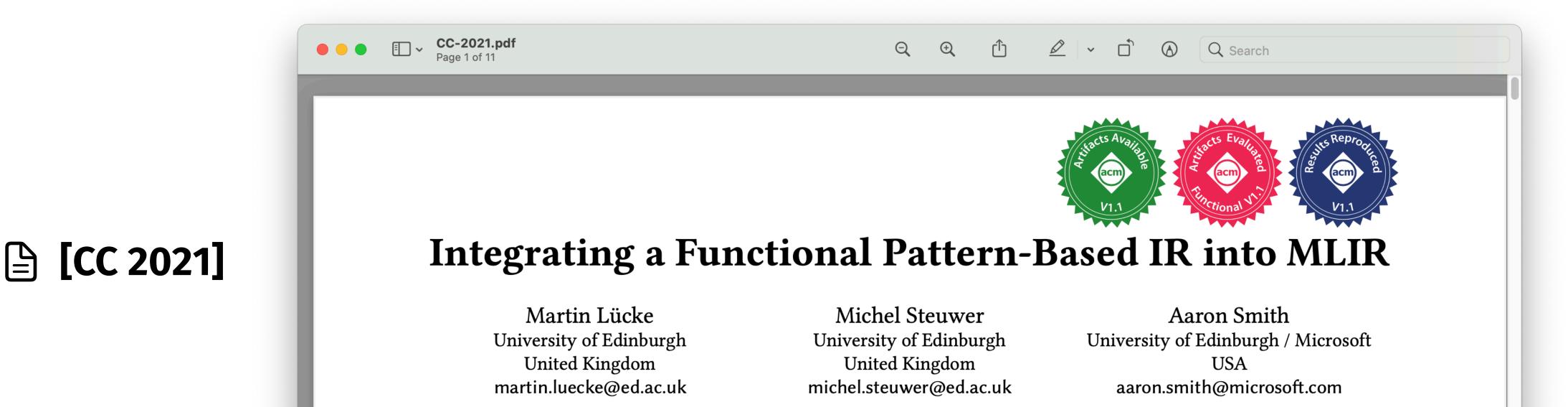
allows to implement state-of-the art scheduling APIs from first principle





MLIR: Offer integration choices

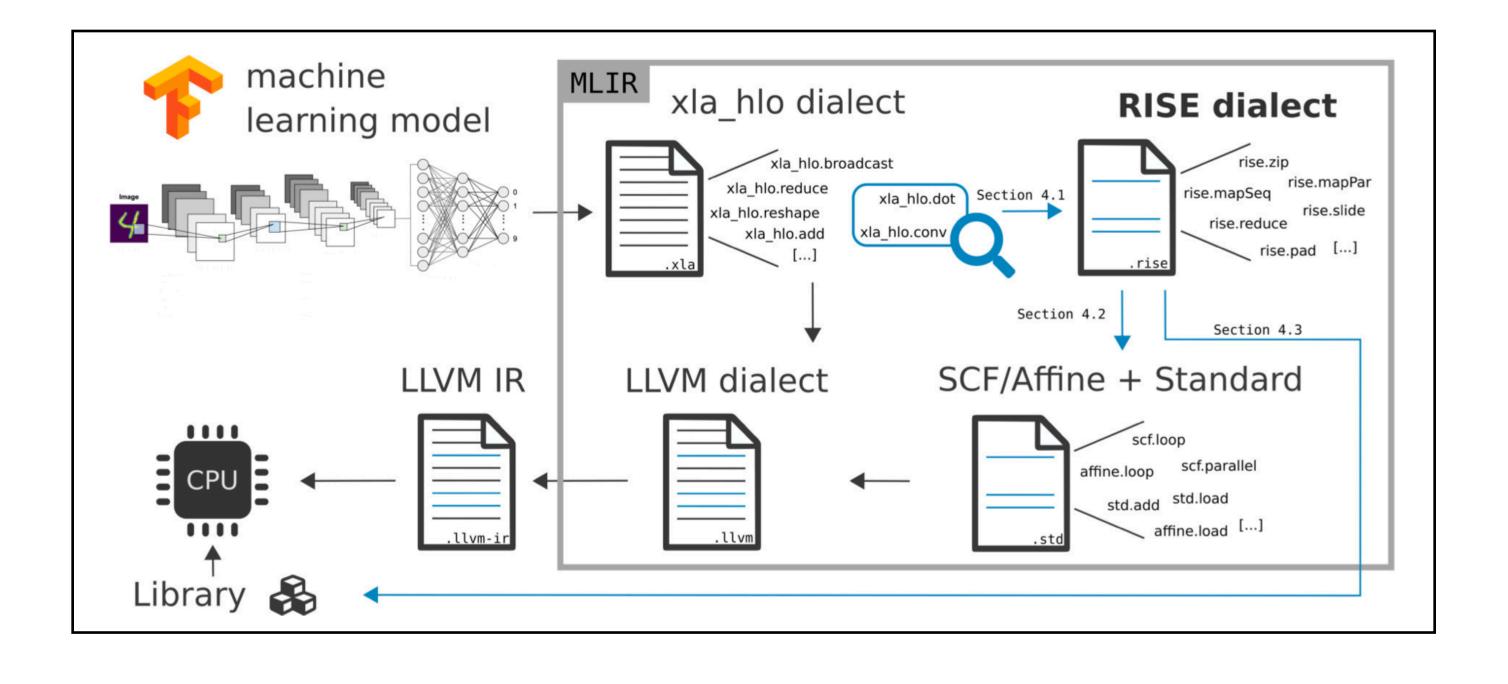
- Focus on compiler intermediate languages rather than user facing languages
 - Avoid flame war over functional programming
 - Type systems (e.g. dependent types) can be complex to carry rich information
 - Easy(er) to build fully integrated systems





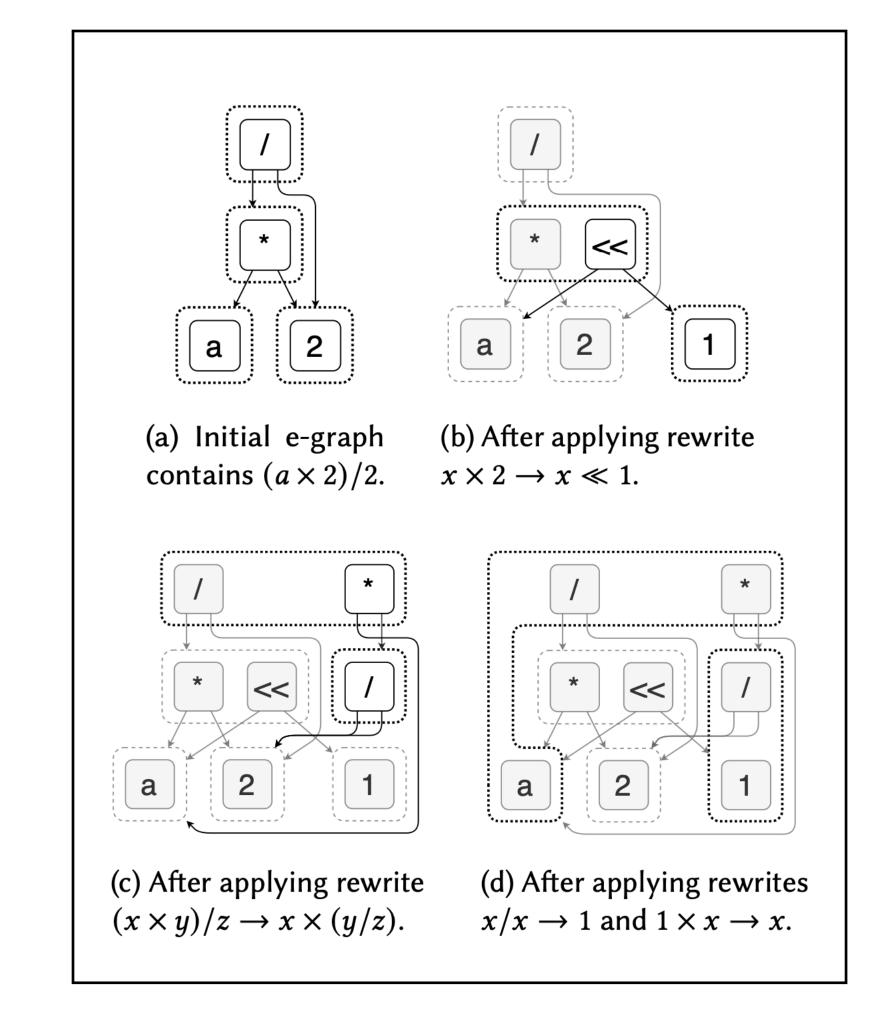
MLIR: Offer integration choices

```
func @mm_fused(%outArg, %inA, %inB) {
        %A = in %inA
        %B = in %inB
        %t = rise.transpose #rise.nat<2048>
                           #rise.nat<2048> #rise.scalar<f32>
        %B_t = rise.apply %t, %B
        %m1fun = lambda (%arow) -> array<2048, scalar<f32>> {
map(fun(bcol,
   ab_1 \times ab_2)+
         %m2fun = lambda (%bcol) -> scalar<f32>
          %zipFun = zip #nat<2048> #scalar<f32> #scalar<f32>
          %zippedArrays = rise.apply %zipFun, %arow, %bcol
          %reduceLambda = lambda(%tuple, %acc)->scalar<f32> 
           %fstFun = rise.fst #scalar<f32> #scalar<f32>
           %sndFun = rise.snd #scalar<f32> #scalar<f32>
           %first = rise.apply %fstFun, %tuple
           %second = rise.apply %sndFun, %tuple
           %result = rise.embed(%first, %second, %acc) {
            %product = mulf %first, %second :f32
transpose
            %result = addf %product, %acc : f32
            return %result : f32
           return %result : scalar<f32>
          %init = rise.literal #lit<0.0>
          %reduceFun = reduceSeq #nat<2048> #tuple
          %result = rise.apply %reduceFun, %reduceLambda,
\mathbf{B}
                                %init, %zippedArrays
          return %result : scalar<f32>
p(fun(arow
         %m2 = mapSeq #nat<2048> #array<2048, scalar<f32>>
                                   #scalar<f32>
         %result = rise.apply %m2, %m2fun, %B_t
         return %result : array<2048, array<2048, scalar<f32>>>
ma
        %m1 = mapSeq #nat<2048> #array<2048, scalar<f32>>
                                       #array<2048, scalar<f32>>
        %result = rise.apply %m1, %m1fun, %A
        out %outArg <- %result</pre>
        return
```



How to make choices?

- Fully manual via ELEVATE
- Fully automated via:
 - Stochastic methods [[ICFP 2015]
 - Equality Saturation & E-graphs:
 - Search "Optimizing Functional Programs with Equality Saturation" on YouTube
 - Reinforcement Learning & other machine learning methods



Big open question: How can we mix both modes conveniently?

Team



Martin Lücke



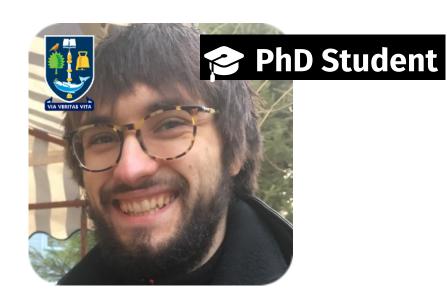
Federico Pizzuti



Xueying Qin



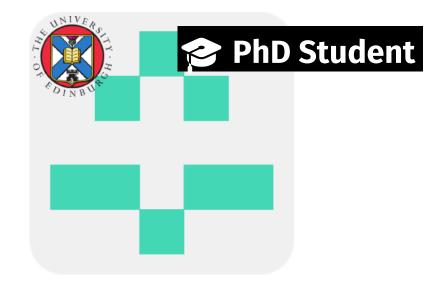
Johannes Lenfers



Thomas Kœhler



Bastian Köpcke



Rongxiao Fu

Compilers

Programming Languages