IRDL: An IR Definition Language for SSA Compilers

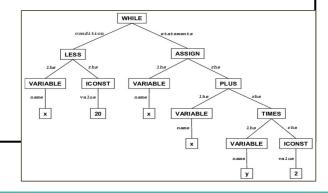
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IR Design for Compilers

• Developers generally use general-purpose programming languages to design IRs. As a result, IR implementations are verbose and manual modifications are expensive.

• While compilers relied historically on a few slowly evolving IRs, domain specific optimizations and specialized hardware motivate compilers to use and evolve many IRs.

```
; Function Attrs: nounwind uwtable
define i32 @main(i32 %argc, i8** %argv) #0 {
entry:
    %retval = alloca i32, align 4
    %argv.addr = alloca i8**, align 8
    %argc.addr = alloca i32, align 4
    store i32 0, i32* %retval, align 4
    store i8** %argv, i8*** %argv.addr, align 8
    store i32 %argc, i32* %argc.addr, align 4
    ret i32 0
}
```



What is IRDL and Why it exists?

• Domain-specific language to define IRs, and it facilitates the implementation of SSA-based IRs.

• It is capable of expressing 28 domain-specific IRs developed as a part of LLVM's MLIR project while only rarely falling back to IRDL's support for generic C++ extensions.

• Aim is to enable concise and explicit specification of IRs and provide foundations for developing effective tooling to automate compiler construction process.

Need for IR design Languages

• LLVM, as a compiler infrastructure, has not only its user-facing LLVM IR but additionally uses various internal ones that are typically not visible to its users. Some of them are SelectionDAG, MachineInst, and MCInst.

• All of these IRs are deeply embedded into their respective compilers. Hence modifications require detailed compiler-specific knowledge, and even specialists are very hesitant to evolve existing IRs.

• While there exists approaches for generating parts of compiler (parsers, backends, code-generators, etc.), we lack a solution that streamlines the design of IRs themselves.

General Properties of IRDL

• MLIR does not provide predefined set of operations but relies on the concept of extensibility, with few built-in constructs leaving most of the IR customizable.

• Operations, types, and attributes are grouped into dialects, similar to namespaces or modular libraries. Each dialect sits at a abstraction level. For example, Linalg dialect in MLIR models linear algebra operations on either tensor or buffer operands.

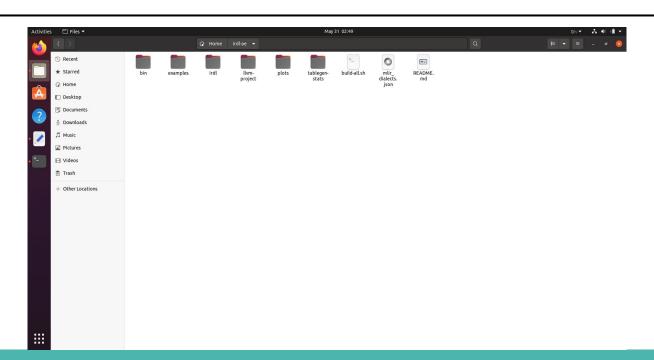
```
func @conorm(%p: !cmath.complex<!f32>, %q: !cmath.complex<!f32>) -> !f32
    %norm_p = cmath.norm(%p) : !f32
    %norme_q = cmath.norm(%q) : !f32
    %pq = std.mulf %norm_p, % norm_q : !f32
    return %pq : !f32
```

Self-Contained IRDL specification

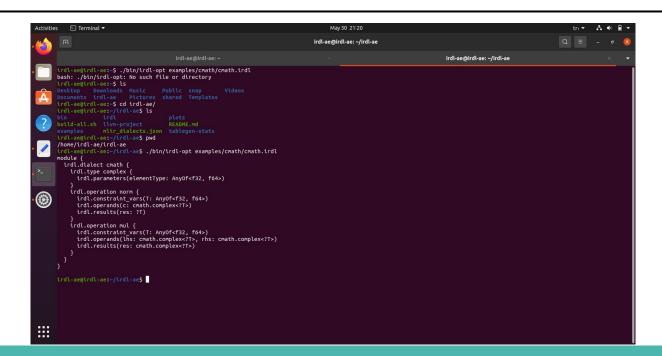
```
Dialect cmath {
     Alias !FloatType = !AnyOf<!f32,!f64>
     Type complex {
          Parameters (elementType: !FloatType)
     Summary "A complex number"
     Operation mul {
       ConstraintVar (!T: !complex<FloatType>)
      Operands (lhs: !T, rhs: !T)
       Results (res: !T)
       Format "$lhs, $rhs: $T.elementType"
      Summary "Multiply two complex numbers"
```

```
Operation norm {
    ConstraintVar (!T: !FloatType)
    Operands(c: !complex<!T>)
    Results (res: !T)
    Format "$c: $T"
    Summary "norm of complex number"
```

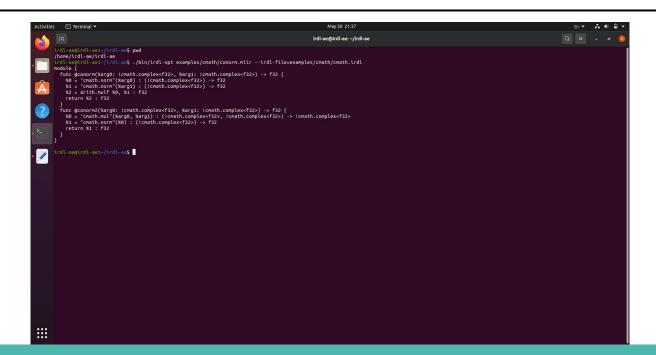
• General layout of the project



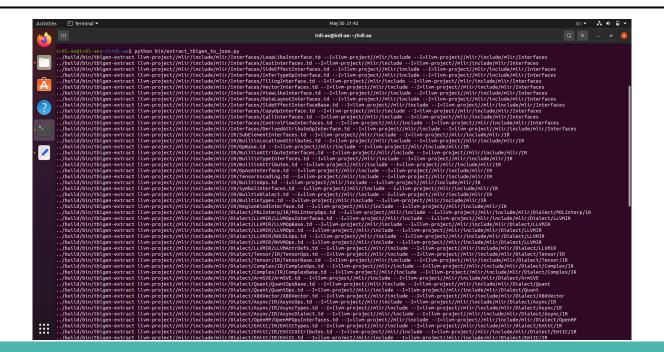
Parsing cmath module with irdl-opt



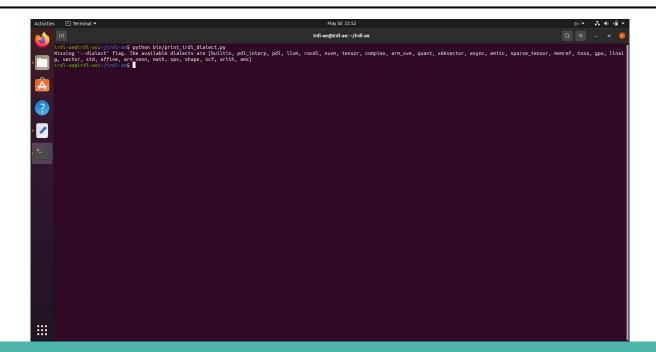
• Registering cmath dialect and parsing the conorm.mlir example



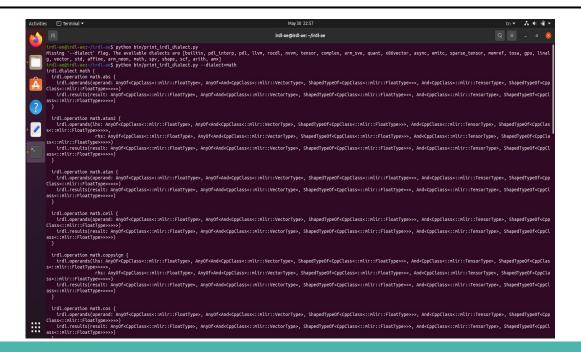
Extracting MLIR dialects in order to convert them into IRDL format



• Querying the list of available dialects



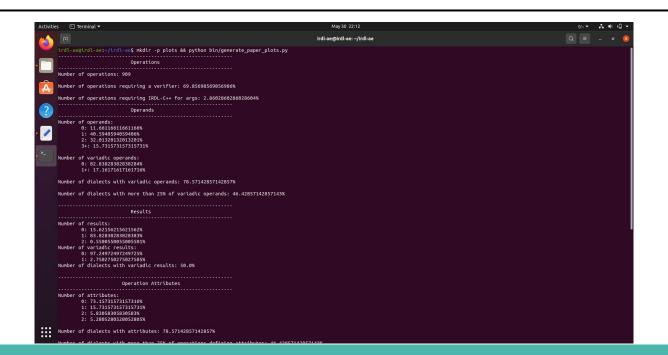
Converting math dialect to IRDL format



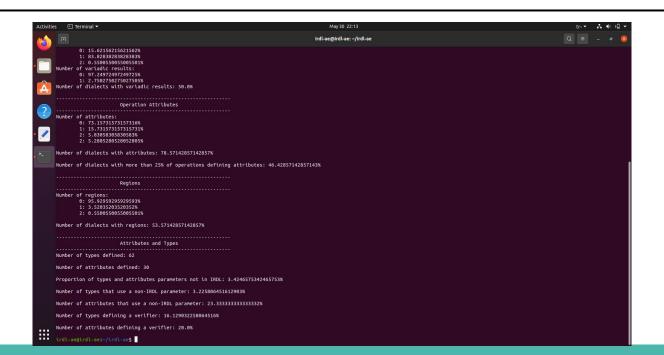
• Converting math affine dialect to IRDL format

```
May 30 22:54
                                                                                                irdl-ae@irdl-ae: ~/irdl-ae
       irdl-ae@irdl-ae:~/irdl-ae$ python bin/print_irdl_dialect.py --dialect=affine
       irdl.operation affine.apply {
          irdl.operands(mapOperands: Variadic<CppClass<::mlir::IndexType>>)
          irdl.results(__empty__: CppClass<::mlir::IndexType>)
         irdl.operation affine.for {
          irdl.operands(__empty__: Variadic<Any>)
          irdl.results(results: Variadic<Any>)
         irdl.operation affine.if {
          irdl.operands(__empty__: Variadic<Any>)
irdl.results(results: Variadic<Any>)
        irdl.operation affine.load {
          irdl.operands(memref: And<CppClass<::mlir::MemRefType>, Any>,
                         indices: Variadic<CopClass<::mlir::IndexTvpe>>)
        irdl.operation affine.max {
          irdl.operands(operands: Variadic<CppClass<::mlir::IndexType>>)
          irdl.results(__empty__: CppClass<::mlir::IndexType>)
        irdl.operation affine.min {
         irdl.operands(operands: Variadic<CppClass<::mlir::IndexType>>)
          irdl.results(_empty_: CppClass<::mlir::IndexType>)
        irdl.operation affine.parallel {
          irdl.operands(mapOperands: Variadic<CppClass<::mlir::IndexType>>)
          irdl.results(results: Variadic<Any>)
        irdl.operation affine.prefetch {
          irdl.operands(memref: And<CppClass<::mlir::MemRefType>, Any>,
                         indices: Variadic<CppClass<::mlir::IndexType>>)
        irdl.operation affine.store {
          irdl.operands(value: Any,
                         memref: And<CppClass<::mlir::MemRefType>, Any>,
                         indices: Variadic<CppClass<::mlir::IndexType>>)
```

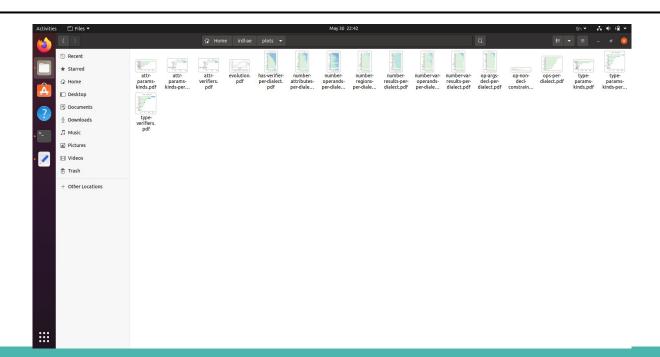
• Reproduced table statistics for Operations, Operands, Results, Operation Attributes



Reproduced table statistics for Regions and Number info for Types and Attributes



• Plots are reproduced with command: mkdir -p plots && python bin/generate_paper_plots.py



References

- https://www.researchgate.net/profile/Peter-Fritzson/publication/228792639/figure/fig
 1/AS:393782852898820@1470896556105/Abstract-syntax-tree-of-the-while-loop.pn
- https://raw.githubusercontent.com/Naios/notepad_llvm/master/preview.png
- https://dl.acm.org/doi/pdf/10.1145/3519939.3523700