

Seismic inversion through operator overloading



Sean Ross-Ross

Synopsis

□ Challenge:

- integrate & scale IO intensive "Pipe"-based software
- does not facilitate transfer of knowledge
- extend flow-based processing to iterative processing

□ Opportunity:

- Create an object-oriented layer
 - Implement algorithms modeled directly from math
 - Independent of the lower-level software.

□ (p)SLIMpy:

- A Collection of Python classes
- vector, linear operator, R/T operation

□ Benefits:

- Reusable
- Scalable (with parallelization including domain decomposition)

Motivation

- Inverse problems in (exploration) seismology are
 - large scale
 - # of unknowns exceeds 2^{30}
 - matrix-free implementation of operators
 - matrix-vector operations take hours, days, weeks
- Software development
 - highly technical coding
 - little code reuse
 - emphasis on processing flows
 - no environment to do iterations as part of optimization

Opportunity

- Create a abstraction layer for the user to implement algorithms
 - Object-oriented
 - Interfaces ANAs (coordinate-free abstract numerical algorithms) with lower level flow-based seismic processing software (e.g. Madagascar)
 - Independent of the lower level software that can be
 - in-core
 - out-of-core (pipe-based)
 - serial or parallel

(p)SLIMpy

- Provides a collection of Python classes and functions, to represent abstract numerical algorithms
- Is a tool used to design and implement algorithms
 - ***clean*** code
 - ***no*** overhead
 - designed to facilitate the **knowledge transfer** between *end users* and the *algorithm designers* (scientists)
- Easily scaleable
 - transparent parallel IO service

(p)SLIMpy

- Industry uses scalable in-core solutions which are not applicable for non-separable transforms in dimensions higher than two (e.g. 3-D curvelet transform)
- SLIM's technology is based on **nonseparable** transforms that take 100.000's of traces as input for a single transform
- SLIMPy is designed to handle large IO intensive non-separable transforms

Context

- SLIMpy is a IO-intensive adaptation of existing ideas:
 - William Symes' Rice Vector Library.
 - <http://www.trip.caam.rice.edu/txt/tripinfo/rvl.html>
 - <http://www.trip.caam.rice.edu/txt/tripinfo/rvl2006.pdf>
 - Ross Bartlett's C++ object-oriented interface, Thyra.
 - <http://trilinos.sandia.gov/packages/thyra/index.html>
 - Reduction/Transformation operators (both part of the Trilinos software package).
 - <http://trilinos.sandia.gov/>
 - PyTrilinos by William Spetz.
 - <http://trilinos.sandia.gov/packages/pytrilinos/>

Methodology

SLIMpy package is divided into four distinct parts

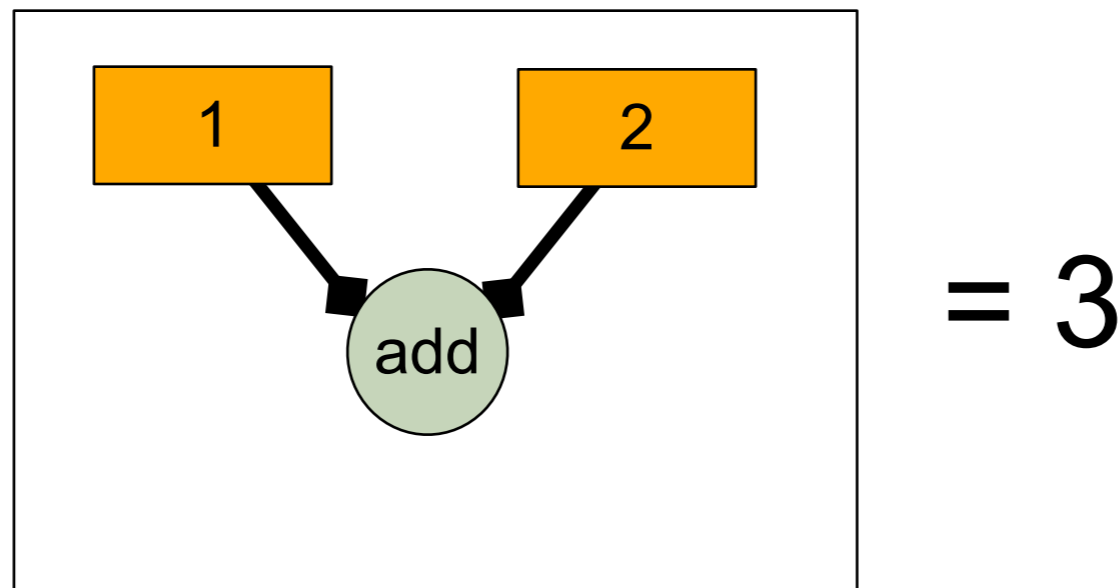
- **Vector/Operator interface**
 - For the algorithm designer
 - Enforce consistency & enable code reuse
- **Abstract Syntax Tree**
 - For SLIMpy developers
 - Allows for pipe optimization & parallelization
- **Compiler**
 - for developers porting to another lower level software
 - Interface to low-level pipe-based commands (e.g. SU, RSF)
- **Reproducible research interface**
 - For end users
 - Allows for access from scons & integration with papers

Operator Overloading

- Operators like $+$, $-$, or $*$ have different implementations depending on the types of their arguments
- SLIMpy uses operator overloading on
 - vectors
 - $\text{vec1} + \text{vec2}$
 - linear operators
 - $\text{Oper} * \text{vec}$
- adds nodes to the Abstract Syntax Tree

Abstract Syntax Tree (AST)

- An AST is a finite, labeled, directed tree where:
 - Internal nodes are labeled by operators
 - Leaf nodes represent variables/vectors



- AST is used as an intermediate between a parse tree and a data structure.

Example

- Apply Operator to two data-sets
- add the transformed data
- apply soft thresholding
- apply adjoint operator

Example SLIMpy

```
vec1 = vector( 'vec1.rsf' )  
vec2 = vector( 'vec2.rsf' )
```

```
C = fft( domain=vec.space )
```

```
coeffs1 = C * vec1
```

```
coeffs2 = C * vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = C.adj() * tmp3
```

```
End()
```

Example SLIMpy

```
vec1 = vector( 'vec1.rsf' )  
vec2 = vector( 'vec2.rsf' )  
C = fft( domain=vec.space )
```



Define Vectors &
Linear Operator

```
coeffs1 = C * vec1
```

```
coeffs2 = C * vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = C.adj() * tmp3
```

```
End()
```

Example SLIMpy

```
vec1 = vector( 'vec1.rsf' )  
vec2 = vector( 'vec2.rsf' )  
  
C = fft( domain=vec.space )
```

```
coeffs1 = C * vec1
```

```
coeffs2 = C * vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = C.adj() * tmp3
```

}

Perform operations
on vectors

```
End()
```

Example SLIMpy

```
vec1 = vector( 'vec1.rsf' )  
vec2 = vector( 'vec2.rsf' )
```

```
C = fft( domain=vec.space )
```

```
coeffs1 = C * vec1
```

```
coeffs2 = C * vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = C.adj() * tmp3
```

```
End()
```

} Compile AST

Example SLIMpy

```
vec1 = vector( 'vec1.rsf' )  
vec2 = vector( 'vec2.rsf' )
```

```
C = fft( domain=vec.space )
```

```
coeffs1 = C * vec1
```

```
coeffs2 = C * vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = C.adj() * tmp3
```

```
End()
```


Highlight data and commands

```
vec1 = vector( 'vec1.rsf' )
```

```
vec2 = vector( 'vec2.rsf' )
```

```
C = fft( domain=vec.space )
```

```
coeffs1 = (C * ) vec1
```

```
coeffs2 = (C * ) vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = (C.adj() * ) tmp3
```

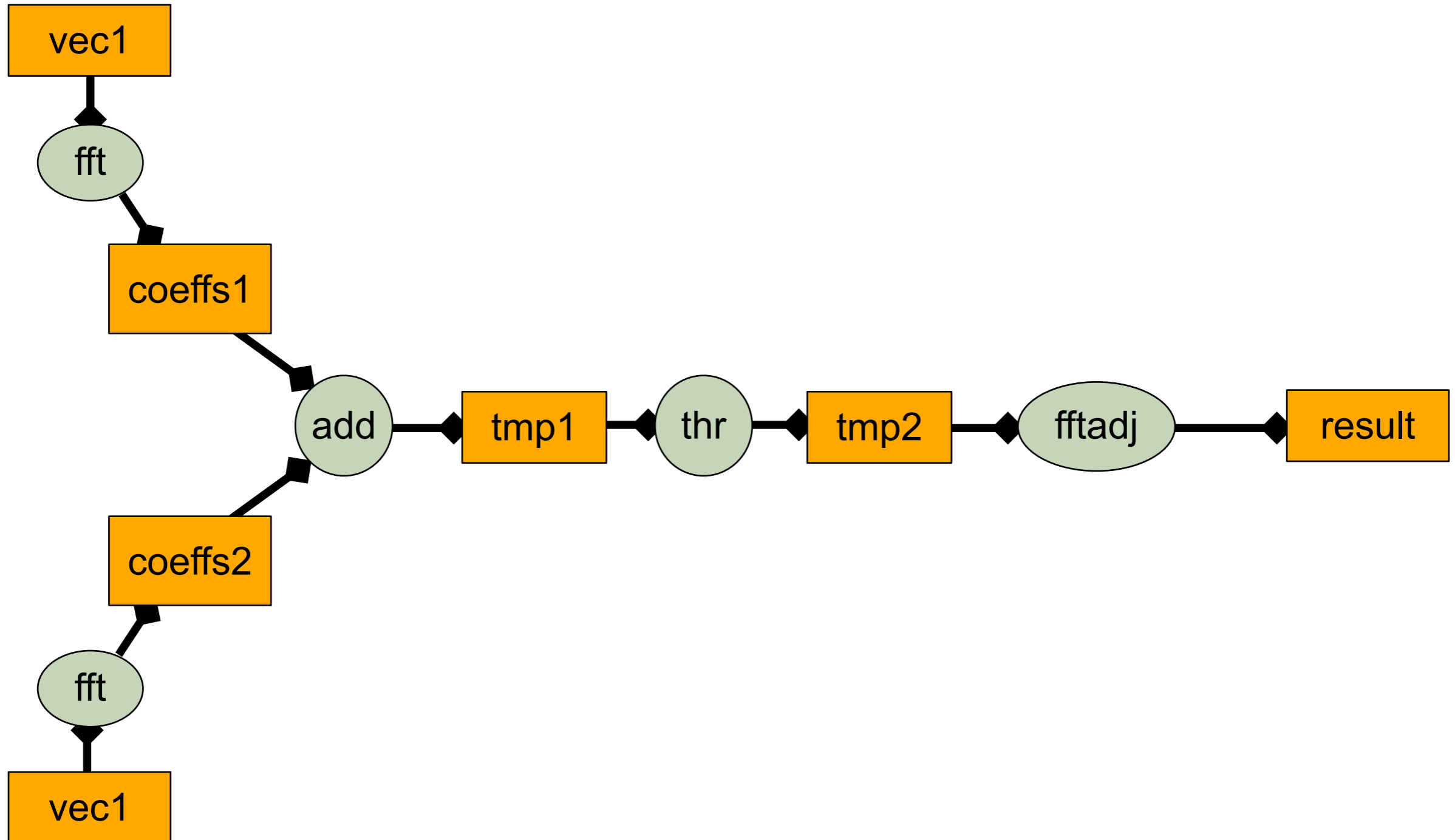
```
End()
```

Can be split into
different nodes:

commands

data

Abstract Syntax Tree



Madagascar

Could work for any Lower Level software

```
< ./data2.rsffft1 opt="n" inv="n" sym="y" | DATAPATH=/Tools/toolboxes/rsf_stuff/tmp_datapath/sffft3 opt="n" inv="n" sym="y" pad="1" axis="2" > /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.12648.env1.FakeMo.fft.
```

00004.rsffft1

```
< ./data1.rsffft1 opt="n" inv="n" sym="y" | sffft3 opt="n" inv="n" sym="y" pad="1" axis="2" | sfmath output="vec +input" vec="/Tools/toolboxes/rsf_stuff/tmp_datapath/slim.12648.env1.FakeMo.fft.00004.rsffft1" | sfthr mode="soft" thr="0.001" | sffft3 opt="n" inv="y" sym="y" pad="1" axis="2" | DATAPATH=/Tools/toolboxes/rsf_stuff/datapath/sffft1 opt="n" inv="y" sym="y" > ./result.rsffft1
```

SLIMpy Compiles

- Most Compilers build an AST
 - c,c++ build and AST with processor instruction set

- SLIMpy builds coarse-grained AST with Linear Algebra commands.
 - reduction/transformation operations that include
 - element-wise addition, subtraction, multiplication
 - vector inner products
 - norms l1, l2 etc
 - Matrix/Vector operations that include
 - implicit linear operators
 - explicit linear operators

Parallelization

- Different Branches of the AST can be run on different hosts.
 - handles copying and organizing data
 - exploits parallel extension of RSF (parallel “file system”)

- SLIMpy can also utilize existing MPI programs

- Look at the previous example again
 - Can use domain decomposition

Parallel: Domain Decomposition

```
vec1 = vector( 'vec1.rsf' )  
vec2 = vector( 'vec2.rsf' )
```

```
P = Scatter( domain=vec.space , [2,1] )  
F = fft( domain=P.range )  
C = CompoundOperator( [F,P] )
```

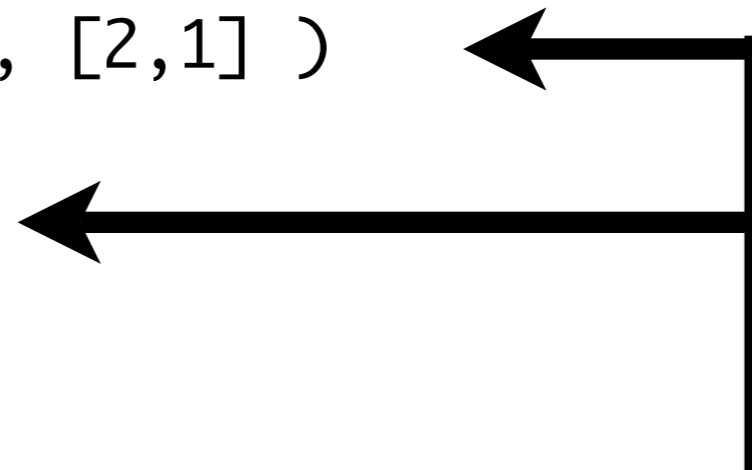
```
coeffs1 = C * vec1  
coeffs2 = C * vec2
```

```
tmp3 = coeffs1 + coeffs2
```

```
tmp4 = tmp3.thr( 0.001 )
```

```
result = C.adj() * tmp3
```

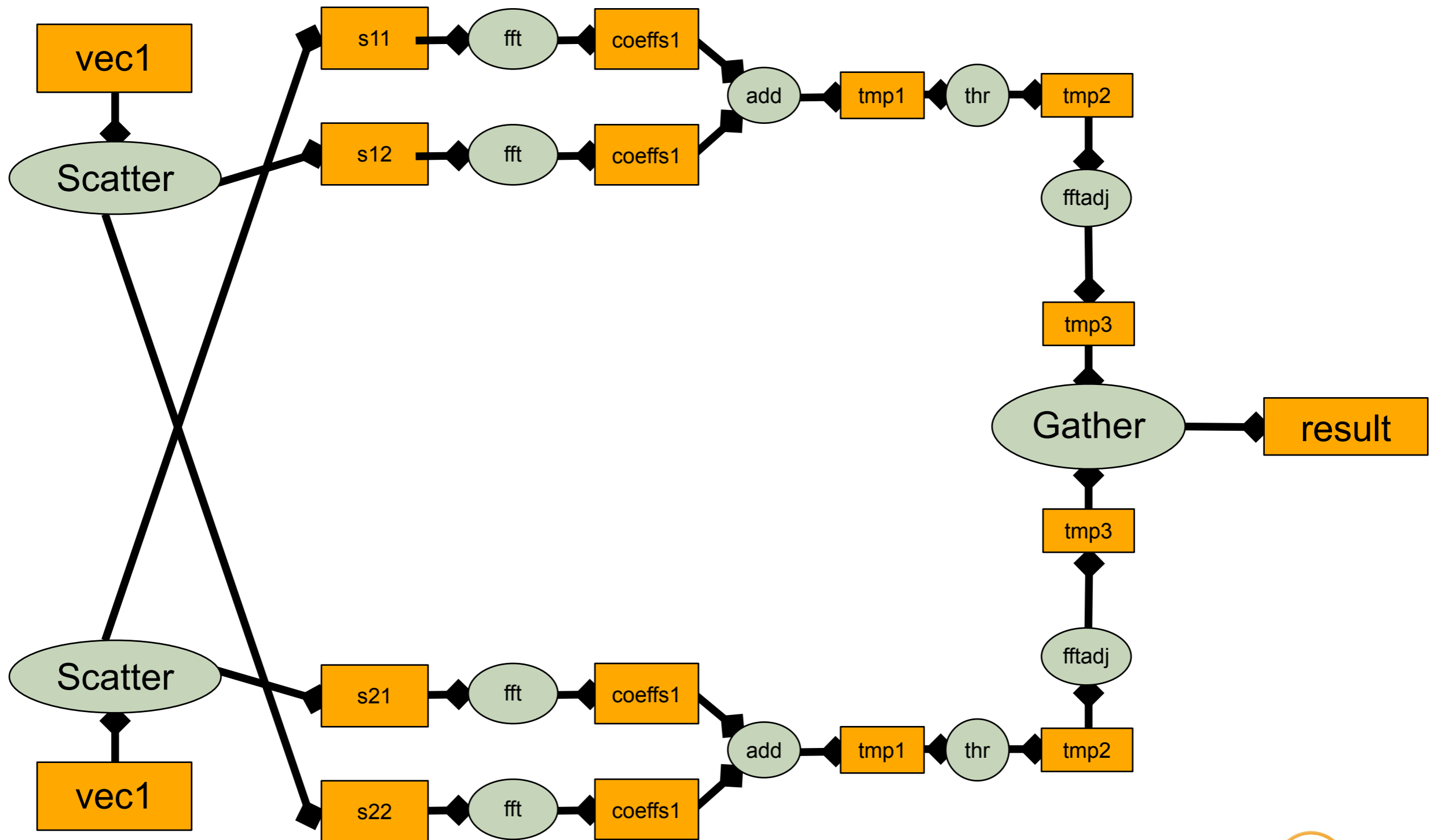
```
End()
```



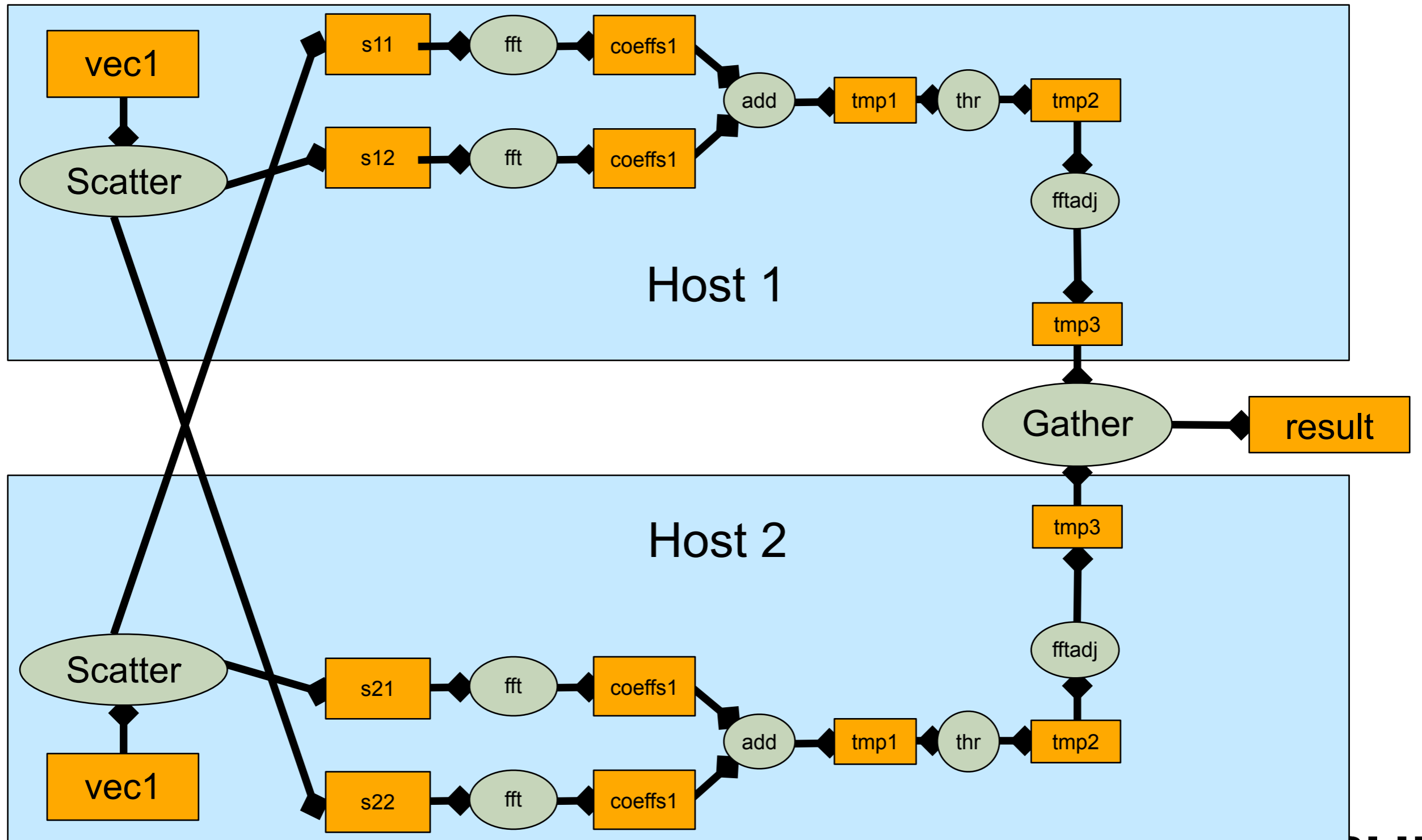
Redefine operator C as:
 $F(P(X))$

C \Leftrightarrow Scatter \rightarrow fft
C.adj() \Leftrightarrow fft inv \rightarrow Gather

Abstract Syntax Tree



Abstract Syntax Tree



Output

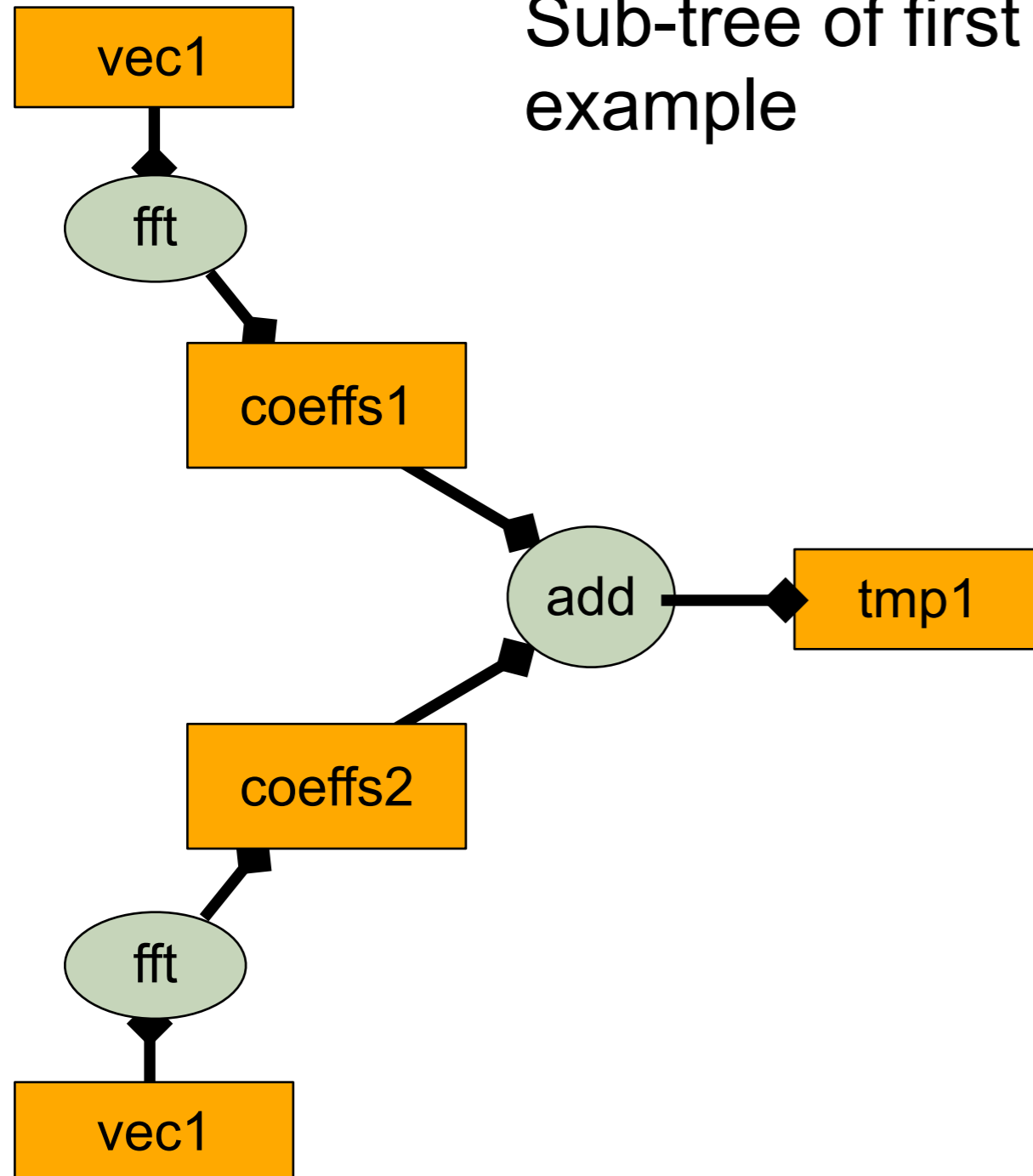
```
< ./s2.rsff sfwindow f1="0" f2="0" n1="5" n2="10" | sffft1 opt="False" inv="n" sym="True" |
DATAPATH=/Tools/toolboxes/rsf_stuff/tmp_datapath/ sffft3 opt="False" inv="n" sym="True" pad="1"
axis="2" > /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft.00011.rsff
< ./s2.rsff sfwindow f1="5" f2="0" n1="5" n2="10" | sffft1 opt="False" inv="n" sym="True" |
DATAPATH=/Tools/toolboxes/rsf_stuff/tmp_datapath/ sffft3 opt="False" inv="n" sym="True" pad="1"
axis="2" > /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft.00012.rsff
< ./s1.rsff sfwindow f1="5" f2="0" n1="5" n2="10" | sffft1 opt="False" inv="n" sym="True" | sffft3
opt="False" inv="n" sym="True" pad="1" axis="2" | sfmath output="vec+input" vec="/Tools/toolboxes/
rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft.00012.rsff" | sfthr mode="soft" thr="0.001" | sffft3
opt="False" inv="y" sym="True" pad="1" axis="2" | DATAPATH=/Tools/toolboxes/rsf_stuff/
tmp_datapath/ sffft1 opt="False" inv="y" sym="True" > /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.
14381.env1.FakeMo.fft1.00020.rsff
< ./s1.rsff sfwindow f1="0" f2="0" n1="5" n2="10" | sffft1 opt="False" inv="n" sym="True" | sffft3
opt="False" inv="n" sym="True" pad="1" axis="2" | sfmath output="vec+input" vec="/Tools/toolboxes/
rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft.00011.rsff" | sfthr mode="soft" thr="0.001" | sffft3
opt="False" inv="y" sym="True" pad="1" axis="2" | sffft1 opt="False" inv="y" sym="True" |
DATAPATH=/Tools/toolboxes/rsf_stuff/datapath/ sfcatt /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.
14381.env1.FakeMo.fft1.00020.rsff axis="1" > ./res.rsff
sfrm /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft1.00020.rsff
sfrm /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft.00012.rsff
sfrm /Tools/toolboxes/rsf_stuff/tmp_datapath/slim.14381.env1.FakeMo.fft.00011.rsff
```

Other Benefits of AST

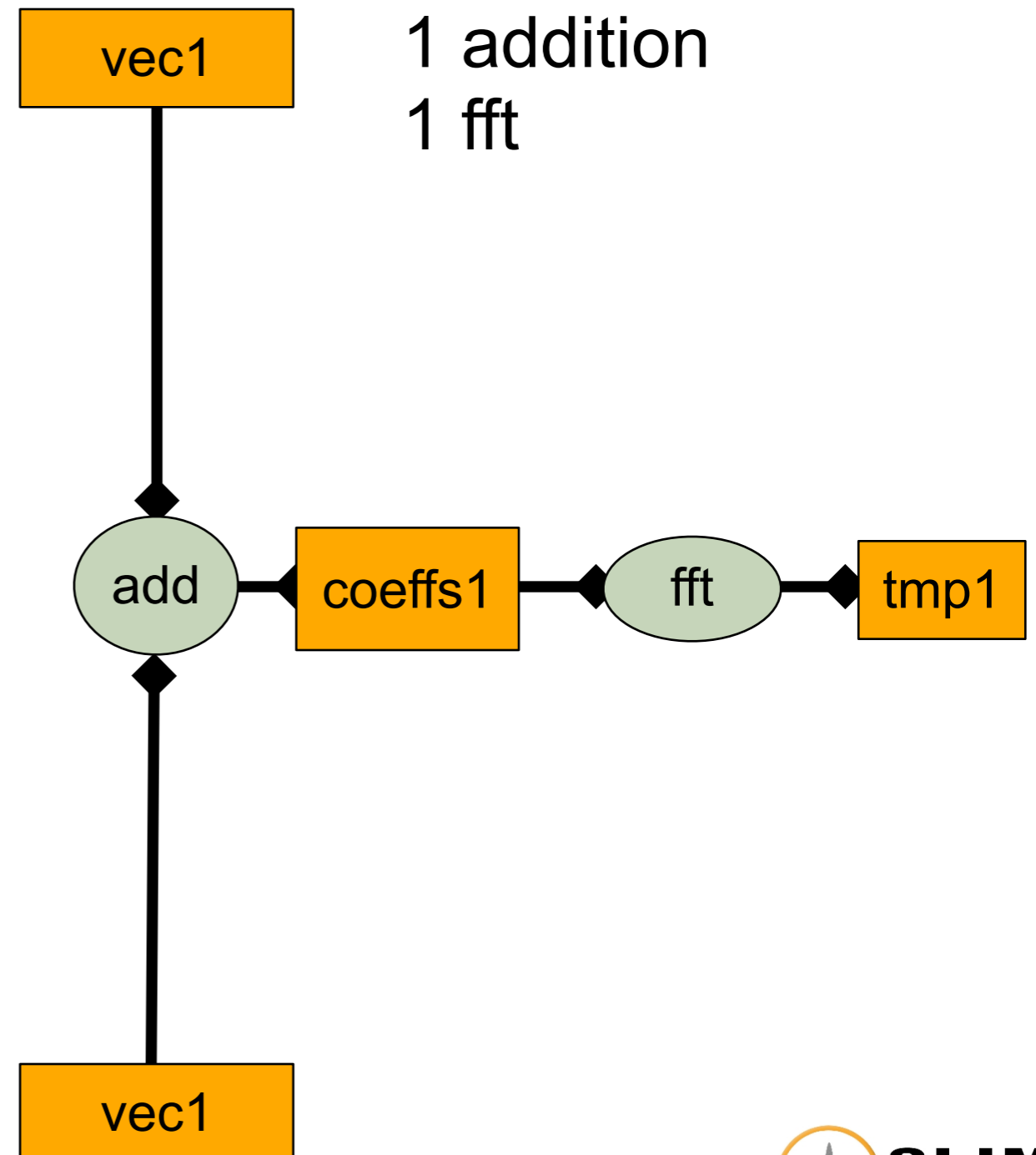
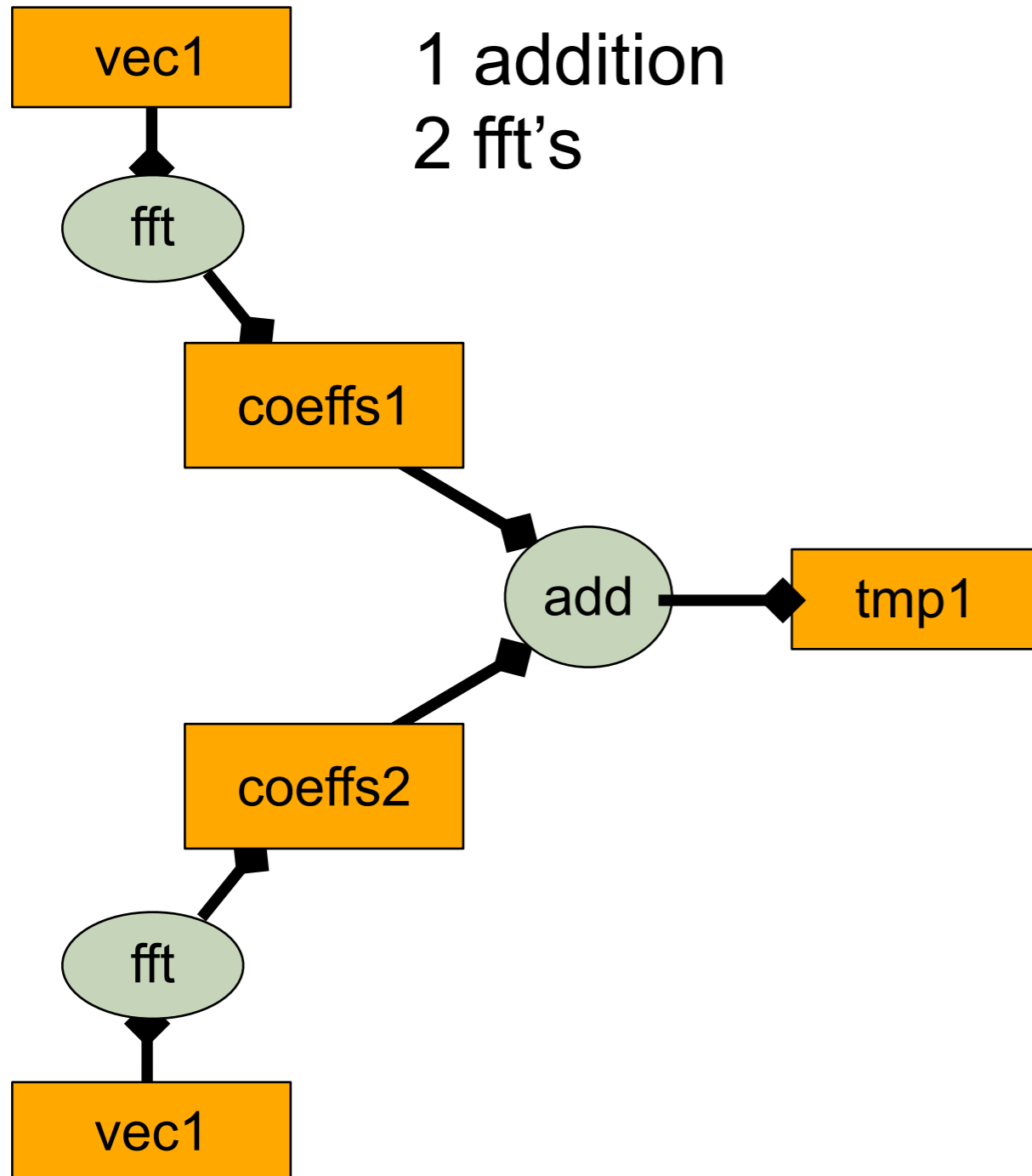
- Lots of existing software use AST
- Potential for optimization
 - C++: has "-O2" flag
- SLIMpy: Symbolic optimization
 - $F(x) + F(y) \Leftrightarrow F(x+y)$
- Useful for augmented system of equations
 - $[F \ F] * [x \ y]^T$

AST Optimization

Sub-tree of first example



AST Optimization



Features

- Automatic Domain/Range checking of Operators
- Automatic dot-test for linear operators
 - $\langle F(x), F(y) \rangle == \langle x, y \rangle$
- Plugin system to add definitions of lower level software

Demo

□ Using an l1-solver

$$\mathbf{P}_\epsilon : \begin{cases} \tilde{\mathbf{x}} = \arg \min_{\mathbf{x}} \|\mathbf{x}\|_1 & \text{s.t.} \quad \|\mathbf{A}\mathbf{x} - \mathbf{y}\|_2 \leq \epsilon \\ \tilde{\mathbf{f}} = \mathbf{S}^T \tilde{\mathbf{x}} \end{cases}$$

□ For 2d and 3d de-noise:

- where $\mathbf{A} = \mathbf{S}^T$ is the inverse curvelet transform

□ For interpolation:

- $\mathbf{A} = [\mathbf{R} \ \mathbf{C}^H]$
- $\mathbf{S}^T = \mathbf{C}^H$

□ For interpolation with parallel and MPI

- $\mathbf{A} = \dots$
- $\mathbf{x} = \dots$

Who should use SLIMpy?

- Anyone working on large-to-extremely large scale optimization:
 - NumPy, Matlab etc. etc.
 - unix pipe-based (Madagascar, SU, SEPlib etc.)
 - seamless migration from in-core to out-of-core to parallel
- Anyone who would like to produce code that is:
 - readable & reusable
 - deployable in different environments
 - integrable with existing OO solver libraries
- *Write solver once, deploy "everywhere" ...*

Conclusions

Use a scripting language to access low-level implementations of (linear) operators (seismic processing tools).

Easy to use automatic checking tools such as domain-range checks and dot-test.

Overloading and abstraction with small overhead.

AST allows for optimization.

Reusable ANAs and Applications.

Is growing into a “compiler” for ANA’s

Future plans

Improve stability of parallel extension

Prepare a public-domain release

Extend the functionality of the AST

- symbolic optimization

Implement the Kronecker product

Acknowledgments

- Madagascar Development Team
 - Sergey Fomel
- CurveLab 2.0.2 Developers
 - Emmanuel Candes, Laurent Demanet, David Donoho and Lexing Ying
- SINBAD project with financial support
 - BG Group, BP, Chevron, ExxonMobil, and Shell
- SINBAD is part of a collaborative research and development grant (CRD) number 334810-05 funded by the Natural Science and Engineering Research Council (NSERC)