

The STCE Scripting Language

Support for Algorithmic Differentiation

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dco/c++

Tangents

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Second-Order Adjoints

- ▶ `dco::gt1s<T>::type` enables the propagation of tangents (directional derivatives) by overloading of all arithmetic operations and intrinsic functions in C++.
- ▶ `dco::ga1s<T>::type` uses overloading to generate a *tape* that records data dependences and elemental partial derivatives. Adjoint are propagated by interpretation of the tape.
- ▶ Recursive nesting of tangent and adjoint types yields higher-order tangents and adjoints.
- ▶ Conceptually, the entire range of functionalities provided by dco/c++ is available for STCE scripting; see
www.nag.com/content/algebraic-differentiation-software/.

dco/c++

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Tangent Algorithmic Differentiation (AD) of differentiable programs implementing multivariate vector functions

$$f : \mathbf{R}^n \rightarrow \mathbf{R}^m : y = f(x)$$

yields the tangent program

$$f^{(1)} : \mathbf{R}^n \times \mathbf{R}^n \rightarrow \mathbf{R}^m \times \mathbf{R}^m : (y, y^{(1)}) = f^{(1)}(x, x^{(1)})$$

defined as

$$f^{(1)} \equiv \left(f(x), f'(x) \cdot x^{(1)} \right).$$

The Jacobian $f' \equiv \frac{df}{dx} \in \mathbf{R}^{m \times n}$ can be computed with a relative (with respect to the cost of evaluating f) cost of $\mathcal{O}(n)$ by letting $x^{(1)}$ range over the Cartesian basis vectors in \mathbf{R}^n .

Tangents with dco/c++

```
1 #include<iostream>
2 using namespace std;
3
4 #include "dco.hpp" // dco/c++ header
5
6 int main() {
7     dco::gt1s<float>::type x,y; // tangent type variables
8     dco::value(x)=2; // set primal value of active input
9     dco::derivative(x)=1; // seed tangent of active input
10    y=sin(x); // evaluate primal and propagate tangent
11    cout << dco::value(y) << " "; // get primal value of active output
12    cout << dco::derivative(y) << endl; // harvest tangent of active output
13    return 0;
14 }
```

yields output

```
0.909297
-0.416147
```

Eigen and dco/c++ can be combined to yield derivatives of linear algebra programs.

For example, differentiation of the solution of the system of linear equations

$$A \cdot x = b ,$$

where $A \in \mathbb{R}^{n \times n}$ is invertible and $x, b \in \mathbb{R}^n$, with respect to b yields

$$\frac{dx}{db} = \frac{d(A^{-1} \cdot b)}{db} = A^{-1}$$

which is easily validated by the following sample code.

Tangents and Eigen

```
1 #include<iostream>
2 using namespace std;
3
4 #include "Eigen.hpp"
5 #include "dco.hpp"
6
7 int main() { // computes the inverse of A as dx/db
8     const int n=3;
9     using T=dco::gt1s<float>::type; // tangent type
10    using VT=Eigen::vector_t<T>; // vector of tangent type variables
11    using MT=Eigen::matrix_t<T>; // matrix of tangent type variables
12    VT x(n), b=VT::Random(n); MT A=MT::Random(n,n);
13    for (int i=0;i<n;i++) {
14        dco::derivative(b(i))=1; // seed
15        x=A.lu().solve(b); // propagate
16        for (int j=0;j<n;j++) cout << dco::derivative(x(j)) << " "; // harvest
17        cout << endl;
18        dco::derivative(b(i))=0; // unseed
19    }
20    cout << A.inverse().transpose() << endl;
21    return 0;
22 }
```

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Adjoint AD of differentiable programs implementing multivariate vector functions

$$f : \mathbf{R}^n \rightarrow \mathbf{R}^m : y = f(x)$$

yields the adjoint program

$$f_{(1)} : \mathbf{R}^n \times \mathbf{R}^m \rightarrow \mathbf{R}^m \times \mathbf{R}^n : (y, x_{(1)}) = f_{(1)}(x, y_{(1)})$$

defined as

$$f_{(1)} \equiv \left(f(x), f'(x)^T \cdot y_{(1)} \right) .$$

The Jacobian can be computed with a relative cost of $\mathcal{O}(m)$ by letting $y_{(1)}$ range over the Cartesian basis vectors in \mathbf{R}^m .

Potentially, the size of the *tape* to be recorded for the necessary reversal of the data flow could exceed the available memory resources. Techniques beyond black-box adjoint AD need to be considered in this case.

```
1 #include<iostream>
2 using namespace std;
3
4 #include "dco.hpp" // dco/c++ header
5
6 int main() {
7     dco::ga1s<float>::type x,y; // adjoint type variables
8     dco::value(x)=2; // set primal value of active input
9     dco::smart_tape_ptr_t<dco::ga1s<float>> tape; // enable recording of tape
10    tape->register_variable(x); // register active input with tape (should be read-only)
11    y=sin(x); // evaluate primal and record tape
12    tape->register_output_variable(y); // register active output (to makes it write-only)
13    cout << dco::value(y) << endl; // get primal value of active input
14    dco::derivative(y)=1; // seed adjoint of active output
15    tape->interpret_adjoint(); // propagate adjoint
16    cout << dco::derivative(x) << endl; // harvest adjoint of active input
17    return 0;
18 }
```

yields the same output as the tangent.

Adjoints and Eigen

```
1 #include<iostream>
2 using namespace std;
3
4 #include "Eigen.hpp"
5 #include "dco.hpp"
6
7 int main() { // computes the inverse of A a dx/db
8     const int n=3;
9     using M=dco::ga1s<float>; using T=M::type;
10    using VT=Eigen::vector_t<T>; using MT=Eigen::matrix_t<T>;
11    VT x(n), b=VT::Random(n); MT A=MT::Random(n,n);
12    dco::smart_tape_ptr_t<M> tape; // enable recording of tape
13    for (int i=0;i<n;i++) tape->register_variable(b(i)); // register active input
14    x=A.lu().solve(b); // record tape
15    for (int i=0;i<n;i++) tape->register_output_variable(x(i)); // register active output
16    for (int i=0;i<n;i++) {
17        dco::derivative(x(i))=1; // seed
18        tape->interpret_adjoint(); // propagate adjoint
19        for (int j=0;j<n;j++) cout << dco::derivative(b(j)) << " ";
20        tape->zero_adoints(); // prepare for re-seed
21    }
22    return 0;
23 }
```

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Tangent AD of a differentiable tangent program

$$f^{(1)} : \mathbf{R}^n \times \mathbf{R}^n \rightarrow \mathbf{R}^m \times \mathbf{R}^m : f^{(1)}(x, x^{(1)}) =: (y, y^{(1)})$$

yields the second-order tangent program

$$\begin{aligned} f^{(1,2)} : & \mathbf{R}^n \times \mathbf{R}^n \times \mathbf{R}^n \times \mathbf{R}^n \rightarrow \mathbf{R}^m \times \mathbf{R}^m \times \mathbf{R}^m \times \mathbf{R}^m \\ & : f^{(1,2)}(x, x^{(2)}, x^{(1)}, x^{(1,2)}) =: (y, y^{(2)}, y^{(1)}, y^{(1,2)}) \end{aligned}$$

defined using index notation for the Hessian tensor product (summation over common index) as

$$f^{(1,2)} \equiv \left(f(x), f'(x) \cdot x^{(2)}, f'(x) \cdot x^{(1)}, f''(x)_{k,j,i} \cdot x_j^{(1)} \cdot x_i^{(2)} + f'(x) \cdot x^{(1,2)} \right).$$

The Hessian $f'' \equiv \frac{d^2 f}{dx^2} \in \mathbf{R}^{m \times n \times n}$ can be computed with a relative cost of $\mathcal{O}(n^2)$ by letting $x^{(1)}$ (first first-order tangent) and $x^{(2)}$ (second first-order tangent) range independently over the Cartesian basis vectors in \mathbf{R}^n while setting $x^{(1,2)} := 0$ (second-order tangent).

```
1 #include<iostream>
2 using namespace std;
3
4 #include "dco.hpp" // dco/c++ header
5
6 int main() {
7     dco::gt1s< dco::gt1s<float>::type>::type x,y; // second-order tangent type variables
8     dco::value(dco::value(x))=2; // set primal value of active input
9     dco::value(dco::derivative(x))=1; // seed first first-order tangent
10    dco::derivative(dco::value(x))=1; // seed second first-order tangent
11    // second-order tangent value of active input is equal to zero by construction
12    y=sin(x); // propagate (primal and tangents)
13    cout << dco::value(dco::value(y)) << " "; // get primal value of active output
14    cout << dco::value(dco::derivative(y)) << " "; // harvest first first-order tangent
15    cout << dco::derivative(dco::value(y)) << " "; // harvest second first-order tangent
16    cout << dco::derivative(dco::derivative(y)) << endl; // harvest second-order tangent
17    return 0;
18 }
```

yields the output 0.909297 -0.416147 -0.416147 -0.909297 similar to ...

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Tangent AD of a differentiable adjoint program¹

$$f_{(1)} : \mathbf{R}^n \times \mathbf{R}^m \rightarrow \mathbf{R}^m \times \mathbf{R}^n : f_{(1)}(x, y_{(1)}) =: (y, x_{(1)})$$

yields the second-order adjoint program

$$\begin{aligned} f_{(1)}^{(2)} : & \mathbf{R}^n \times \mathbf{R}^n \times \mathbf{R}^m \times \mathbf{R}^m \rightarrow \mathbf{R}^m \times \mathbf{R}^m \times \mathbf{R}^n \times \mathbf{R}^n \\ & : f_{(1)}^{(2)}(x, x^{(2)}, y_{(1)}, y_{(1)}^{(2)}) =: (y, y^{(2)}, x_{(1)}, x_{(1)}^{(2)}) \end{aligned}$$

defined as

$$f_{(1)}^{(2)} \equiv \left(f(x), f'(x) \cdot x^{(2)}, f'(x)^T \cdot y_{(1)}, f''(x)_{k,j,i} \cdot y_{(1)_k} \cdot x_i^{(2)} + f'(x)^T \cdot y_{(1)}^{(2)} \right).$$

The Hessian can be computed with a relative cost of $\mathcal{O}(m \cdot n)$ by letting $y_{(1)}$ (first-order adjoint) and $x^{(2)}$ (first-order tangent) range independently over the Cartesian basis vectors in \mathbf{R}^m and \mathbf{R}^n , respectively, while setting $y_{(1)}^{(2)} := 0$ (second-order adjoint).

¹The remaining two combinations work too.

Second-Order Adjoints with dco/c++

```
1 #include<iostream>
2 using namespace std;
3
4 #include "dco.hpp" // dco/c++ header
5
6 int main() {
7     using M=dco::ga1s<dco::gt1s<float>::type>; // second-order adjoint mode
8     M::type x,y; // second-order adjoint type variables
9     dco::value(dco::value(x))=2; // set primal value of active input
10    dco::derivative(dco::value(x))=1; // seed first-order tangent
11    dco::smart_tape_ptr_t<M> tape; // enable recording of tape
12    tape->register_variable(x); // register active input
13    y=sin(x); // evaluate primal and record tape
14    tape->register_output_variable(y); // register active output
15    cout << dco::value(dco::value(y)) << " "; // get primal value of active output
16    cout << dco::derivative(dco::value(y)) << " "; // harvest first-order tangent
17    dco::derivative(y)=1; // seed first-order adjoint of active output
18    tape->interpret_adjoint(); // propagate adjoint
19    cout << dco::value(dco::derivative(x)) << " "; // harvest first-order adjoint
20    cout << dco::derivative(dco::derivative(x)) << endl; // harvest second-order adjoint
21    return 0;
22 }
```

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