The Basic Idea

Vector function in $\mathbb{C}/\mathbb{C}++$: $F:\mathbb{R}^n \to \mathbb{R}^m: x \mapsto y = F(x)$

 \downarrow Operator overloading (C++)

Internal representation of F ($\equiv tape$)



Interpretation



Forward mode

$$x\left(t\right) = \sum_{j=0}^{d} x_j t^j$$

$$y(t) = \sum_{j=0}^{d} y_j t^j + O(t^{d+1})$$

 \Rightarrow Directional derivatives $||\Longrightarrow$ Gradients (adjoints)

Reverse mode

$$x(t) = \sum_{j=0}^{d} x_{j} t^{j}$$

$$\downarrow \qquad \qquad \downarrow$$

$$y_{j} = y_{j} (x_{0}, x_{1}, \dots, x_{j})$$

$$\downarrow \qquad \qquad \downarrow$$

$$\frac{\partial y_{j}}{\partial x_{i}} = \frac{\partial y_{j-i}}{\partial x_{0}}$$

$$= A_{j-i} (x_{0}, x_{1}, \dots, x_{j-i})$$

$$y_{0} = F(x_{0})$$

$$y_{1} = F'(x_{0}) x_{1}$$

$$y_{2} = F'(x_{0}) x_{2} + \frac{1}{2} F''(x_{0}) x_{1} x_{1}$$

$$y_{3} = F'(x_{0}) x_{3} + F''(x_{0}) x_{1} x_{2} + \frac{1}{6} F'''(x_{0}) x_{1} x_{1} x_{1}$$

$$\begin{split} \frac{\partial y_0}{\partial x_0} &= \frac{\partial y_1}{\partial x_1} = \frac{\partial y_2}{\partial x_2} = \frac{\partial y_3}{\partial x_3} = A_0 = F'\left(x_0\right) \\ &\frac{\partial y_1}{\partial x_0} = \frac{\partial y_2}{\partial x_1} = \frac{\partial y_3}{\partial x_2} = A_1 = F''\left(x_0\right) x_1 \\ &\frac{\partial y_2}{\partial x_0} = \frac{\partial y_3}{\partial x_1} = A_2 = F''\left(x_0\right) x_2 + \frac{1}{2} F'''\left(x_0\right) x_1 x_1 \\ &\frac{\partial y_3}{\partial x_0} = A_3 = F''\left(x_0\right) x_3 + F'''\left(x_0\right) x_1 x_2 + \frac{1}{6} F^{(4)}\left(x_0\right) x_1 x_1 x_1 \end{split}$$

Application

Operator overloading concept \Rightarrow Code modification

- Inclusion of appropriate ADOL-C headers
- Retyping of all involved variables to active data type adouble
- Marking active section to be "taped" (trace_on/trace_off)
- Specification of independent and dependent variables (<<=/>>=)
- Specification of differentiation task(s)
- Recompilation and Linking with ADOL-C library libad.a

Example:

```
#include "adolc.h"
                                        // inlusion of ADOL-C headers
adouble foo ( adouble x )
                                        // some activated function
{ adouble tmp;
  tmp = log(x);
 return 3.0*tmp*tmp + 2.0;
}
int main (int argc, char* argv[])
                                        // main program or other procedure
 double
         x[2], y;
                                        // declaration of active variables
  adouble ax[2], ay;
  x[0]=0.3; x[1]=2.3;
  trace_on(1);
                                        // starting active section
    ax[0] <<=x[0]; ax[1] <<=x[1];
                                        // marking independent variables
   ay=ax[0]*sin(ax[1])+foo(ax[1]);
                                        // function evaluation
    ay >> = y;
                                         // marking dependend variables
  trace_off();
                                         // ending active section
  double g[2];
  gradient(1,2,x,g);
                                        // application of ADOL-C routine
  x[0]+=0.1; x[1]+=0.3;
                                        // application at different argument
  gradient(1,2,x,g);
}
```

Drivers for Optimization and Nonlinear Equations (C/C++)

$$\min_{x} f(x), \qquad f: \mathbb{R}^{n} \to \mathbb{R}$$
$$F(x) = 0_{m}, \qquad F: \mathbb{R}^{n} \to \mathbb{R}^{m}$$

<pre>function(tag,m,n,x[n],y[m])</pre>	$F\left(x_{0}\right)$
<pre>gradient(tag,n,x[n],g[n]) hessian(tag,n,x[n],H[n][n])</pre>	$\nabla f(x_0) \\ \nabla^2 f(x_0)$
<pre>jacobian(tag,m,n,x[n],J[m][n]) vec_jac(tag,m,n,repeat?,x[n],u[m],z[n]) jac_vec(tag,m,n,x[n],v[n],z[m])</pre>	$F'(x_0)$ $u^T F'(x_0)$ $F'(x_0) v$
hess_vec(tag,n,x[n],v[n],z[n]) lagra_hess_vec(tag,m,n,x[n],v[n],u[m],h[n])	$\nabla^2 f(x_0) v$ $u^T F''(x_0) v$
<pre>jac_solv(tag,n,x[n],b[n],sparse?,mode?)</pre>	$F'(x_0) w = b$

```
Solution of F(x) = 0 by Newton's method
Example:
double x[n], r[n];
int i;
initialize(x);
                                        // setting up the initial x
function(ftag,n,n,x,r);
                                        // compute residuum r
while (norm(r) > EPSILON)
                                       // terminate if small residuum
                                       // compute r:=F'(x)^(-1)*r
{ jac_solv(ftag,n,x,r,0,2);
  for (i=0; i<n; i++)
                                        // update x
    x[i] = r[i];
  function(ftag,n,n,x,r);
                                        // compute residuum r
}
```

Lowest-level Differentiation Routines

 $F: \mathbb{R}^n \to \mathbb{R}^m$

Forward Mode (C/C++)

zos_forward(tag,m,n,keep,x[n],y[m])

- zero-order scalar forward; computes y = F(x)
- $0 \le \text{keep} \le 1$; keep = 1 prepares for fos_reverse or fov_reverse

fos_forward(tag,m,n,keep,x0[n],x1[n],y0[m],y1[m])

- first-order scalar forward; computes $y_0 = F(x_0), y_1 = F'(x_0) x_1$
- $0 \le \text{keep} \le 2$; $\text{keep} = \begin{cases} 1 & \text{prepares for fos_reverse or fov_reverse} \\ 2 & \text{prepares for hos_reverse or hov_reverse} \end{cases}$

fov_forward(tag,m,n,p,x[n],X[n][p],y[m],Y[m][p])

• first-order vector forward; computes y = F(x), Y = F'(x) X

 $hos_forward(tag,m,n,d,keep,x[n],X[n][d],y[m],Y[m][d])$

- higher-order scalar forward; computes $y_0 = F(x_0)$, $y_1 = F'(x_0) x_1$, ..., where $x = x_0$, $X = [x_1, x_2, ..., x_d]$ and $y = y_0$, $Y = [y_1, y_2, ..., y_d]$
- $0 \le \text{keep} \le d+1$; keep $\begin{cases} = 1 \text{ prepares for fos_reverse or fov_reverse} \\ > 1 \text{ prepares for hos_reverse or hov_reverse} \end{cases}$

hov_forward(tag,m,n,d,p,x[n],X[n][p][d],y[m],Y[m][p][d])

• higher-order vector forward; computes $y_0 = F(x_0), Y_1 = F'(x_0)X_1, \ldots,$ where $x = x_0, X = [X_1, X_2, \ldots, X_d]$ and $y = y_0, Y = [Y_1, Y_2, \ldots, Y_d]$

Reverse Mode (C/C++)

fos_reverse(tag,m,n,u[m],z[n])

- first-order scalar reverse; computes $z^{T} = u^{T}F'(x)$
- after calling zos_forward, fos_forward, or hos_forward with keep = 1

fov_reverse(tag,m,n,q,U[q][m],Z[q][n])

- first-order vector reverse; computes Z = UF'(x)
- after calling zos_forward, fos_forward, or hos_forward with keep = 1

hos_reverse(tag,m,n,d,u[m],Z[n][d+1])

- higher-order scalar reverse; computes the adjoints $z_0^T = u^T F'(x_0) = u^T A_0$, $z_1^T = u^T F''(x_0) x_1 = u^T A_1, \ldots$, where $Z = [z_0, z_1, \ldots, z_d]$
- after calling fos_forward or hos_forward with keep = d + 1 > 1

hov_reverse(tag,m,n,d,q,U[q][m],Z[q][n][d+1],nz[q][n])

- higher-order vector reverse; computes the adjoints $Z_0 = UF'(x_0) = UA_0$, $Z_1 = UF''(x_0) x_1 = UA_1, \ldots$, where $Z = [Z_0, Z_1, \ldots, Z_d]$
- after calling fos_forward or hos_forward with keep = d+1>1
- optional nonzero pattern nz (⇒ manual)

Example:

Low-level Differentiation Routines

Forward Mode (C++ interfaces)

forward(tag,m,n,d,keep,X[n][d+1],Y[m][d+1]) forward(tag,m=1,n,d,keep,X[n][d+1],Y[d+1])	hos, fos, zos hos, fos, zos
<pre>forward(tag,m,n,d=0,keep,x[n],y[m]) forward(tag,m,n,keep,x[n],y[m])</pre>	zos
forward(tag,m,n,p,x[n],X[n][p],y[m],Y[m][p])	fov
forward(tag,m,n,d,p,x[n],X[n][p][d], y[m],Y[m][p][d])	hov

Reverse Mode (C++ interfaces)

reverse(tag,m,n,d,u[m],Z[n][d+1]) forward(tag,m=1,n,d,u,Z[n][d+1])	hos
reverse(tag,m,n,d=0,u[m],z[n]) reverse(tag,m=1,n,d=0,u,z[n])	fos
	hov hov
<pre>reverse(tag,m,n,d=0,q,U[q][m],Z[q][n]) reverse(tag,m,n,q,U[q][m],Z[q][n] reverse(tag,m=1,n,d=0,q,U[q],Z[q][n])</pre>	fov fov

Drivers for Ordinary Differential Equations (C/C++)

ODE:
$$x'(t) = y(t) = F(x(t)), \quad x(0) = x_0$$

forodec(tag,n,tau,dold,d,X[n][d+1])

- recursive forward computation of $x_{d_{old}+1}, \ldots, x_d$ from $x_0, \ldots, x_{d_{old}}$ (by $x_{i+1} = \frac{1}{1+i}y_i$)
- application with $d_{old} = 0$ delivers truncated Taylor series $\sum_{i=0}^{d} x_{i} t^{j}$ at base point x_{0}

hov_reverse(tag,n,n,d-1,n,I[n][n],A[n][n][d],nz[n][n])

- reverse computation of $A_j = \frac{\partial y_j}{\partial x_0}, j = 0, \dots, d$ after calling forodec with degree d
- optional nonzero pattern nz (⇒ manual)

accodec(n,tau,d-1,A[n][n][d],B[n][n][d],nz[n][n])

- accumulation of total derivatives $B_j = \frac{dx_j}{dx_0}$, $j = 0, \ldots, d$ from the partial derivatives $A_j = \frac{\partial y_j}{\partial x_0}$, $j = 0, \ldots, d$ after calling hov_reverse
- optional nonzero pattern nz (⇒ manual)

C++: Special C++ interfaces can be found in file SRC/DRIVERS/odedrivers.h!

Example:

ADOL-C provides

- Low-level differentiation routines (forward/reverse)
- Easy-to-use driver routines for
 - the solution of optimization problems and nonlinear equations
 - the integration of ordinary differential equations
 - the evaluation of higher derivative tensors $(\Rightarrow \text{manual})$
- Derivatives of implicit and inverse functions $(\Rightarrow \text{manual})$
- Forward and backward dependence analysis (⇒ manual)

Recent developments

- Efficient detection of Jacobian/Hessian sparsity structure
- Exploitation of Jacobian/Hessian sparsity by matrix compression
- Integration of checkpointing routines
- Exploitation of fixpoint iterations
- Differentiation of OpenMP parallel programs

Future developments

- Internal optimizations to reduce storage needed for reverse mode
- Recovery of structure for internal function representation
- Differentiation of MPI parallel programs

Contact/Resources

• E-mail: adol-c@list.coin-or.org

• WWW: http://www.coin-or.org/projects/ADOL-C.xml