

# Differentiating GATE/Geant4 with Derivgrind

Max Ahle

Scientific Computing Group,  
University of Kaiserslautern-Landau  
& SIVERT Research Training Group

Workshop on Differentiable and  
Probabilistic Programming for Fundamental Physics

MIAPbP Munich, Germany  
June 16<sup>th</sup>, 2023

# Algorithmic Differentiation

- Given a computer-implemented function, AD provides computer-implemented derivative functions, e. g. for gradient-based optimization.
- Based on the idea to insert **AD logic** for all real-arithmetic calculations performed by the primal program.
- **Forward-mode AD propagates dot values** alongside the primal program execution.
- **Reverse-mode AD** back-propagates adjoint values in a reverse pass, after a forward-pass **recording**.

## Ways to Add AD Logic

Classification of AD tools according to mechanism used for instrumentation:

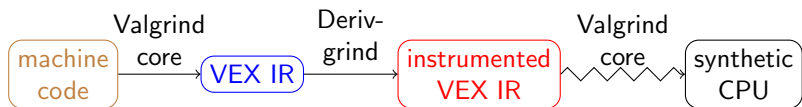
- Source Transformation: TAPENADE, ...
- Operator Overloading: ADOL-C, autograd, CoDiPack, ...
- Compiler-Based: CLAD, Enzyme, ...
- **Machine Code-Based: Derivgrind**

Machine code-based AD reduces interaction with the primal program's source code to the bare minimum, which is desirable if

- the program is written in a (combination of) programming languages for which there is no suitable source code-based AD tool, or
- libraries used by the program are only available in compiled form, or
- to reduce the effort of integrating an AD tool into a primal program, e. g. for exploratory studies.

## Valgrind and VEX IR

We implemented Derivgrind as a “tool” in the Valgrind instrumentation framework for building dynamic analysis tools.



E. g., the x86 instruction `mulsd %xmm0, %xmm1` to multiply scalar double-precision value in register `%xmm0` to `%xmm1` could be translated as

```
t0 = 0x0:I32
```

```
t1 = GET:F64(160)
```

```
t2 = GET:F64(176)
```

```
t3 = MulF64(t0,t1,t2)
```

```
PUT(176) = t3
```

## Forward Mode: Propagation of Dot Values

For every number  $a$ , keep track of  $\dot{a} = \frac{\partial a}{\partial x}$  for single AD input  $x$ .

- Store dot value of temporary  $t\langle i \rangle$  in shadow temporary  $t\langle i + m_{\text{tmp}} \rangle$ .
- Store dot value of register with byte offset  $j$  in shadow register at byte offset  $j + m_{\text{gs}}$ .
- Store dot value of memory using a shadow memory tool.
- The dot value of a constant is zero.
- Use differentiation rules like

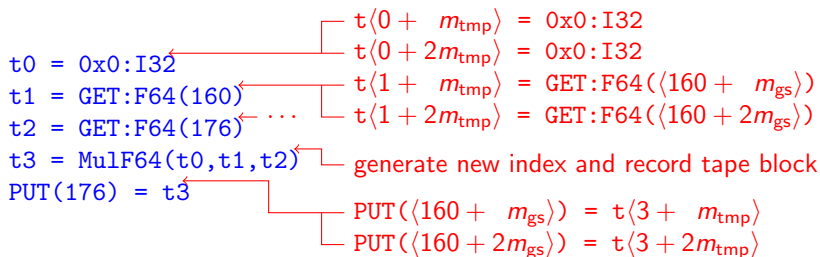
$$a_3 = a_1 \cdot a_2 \rightsquigarrow \dot{a}_3 = \dot{a}_1 \cdot a_2 + a_1 \cdot \dot{a}_2.$$

<code>t0 = 0x0:I32</code>	←	<code>t⟨0 + m<sub>tmp</sub>⟩ = 0x0:I32</code>
<code>t1 = GET:F64(160)</code>	←	<code>t⟨1 + m<sub>tmp</sub>⟩ = GET:F64(⟨160 + m<sub>gs</sub>⟩)</code>
<code>t2 = GET:F64(176)</code>	←	<code>t⟨2 + m<sub>tmp</sub>⟩ = GET:F64(⟨176 + m<sub>gs</sub>⟩)</code>
<code>t3 = MulF64(t0,t1,t2)</code>	←	<code>t⟨3 + m<sub>tmp</sub>⟩ = AddF64(t0,</code>
<code>PUT(176) = t3</code>	←	<code>MulF64(t0,t⟨1 + m<sub>tmp</sub>⟩,t2),</code>
		<code>MulF64(t0,t1,t⟨2 + m<sub>tmp</sub>⟩) )</code>
	←	<code>PUT(⟨160 + m<sub>gs</sub>⟩) = t⟨3 + m<sub>tmp</sub>⟩</code>

## Reverse Mode: Tape Recording

For every number  $a$ , identify it by an *index* and record a *tape*

- Store indices in two layers of shadow memory.
- Passive variables have the default index 0.
- Linear index management and Jacobian tape.



# Bit-Tricks

## Supported “bit-tricks”:

- Compute the negative value of a floating-point number by flipping the sign bit with a bitwise logical “xor”.
- ...

## Unsupported “bit-tricks”:

- Double a number by incrementing the exponent bits.
- ... ..
- ↪ Assumed not to be present in the primal program. (OK in libm.)
- ↪ (Would also be problematic for source-code-based AD tools, to a lesser extent.)

## Specifying Input and Output Variables

In order to identify input and output variables in the compiled program, limited access to the primal program's source code is necessary.

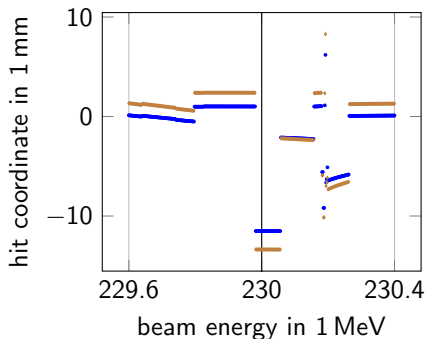
**Client requests** interface: Call specific C macros/functions from the primal code. Requires write access and recompilation of specific source files.

We will now consider the following example:  
Derivgrind's forward mode is applied to GATE/Geant4.



## Application to GATE/Geant4

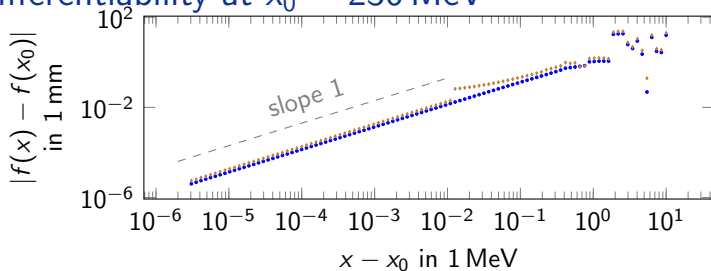
- Monte Carlo simulation for particle physics,  $> 1$  million lines of C++ code.
- In our setup, a single energetic proton passes through a human head and a digital tracking calorimeter. Our input is the beam energy, and our output is a hit coordinate in the first two tracking layers.
- The dependency is piecewise differentiable in an interval around 230 MeV.



- First tracking layer.
- ♦ Second tracking layer.

Adapted from Aehle, Alme et al.,  
Derivatives in Proton CT [↗](#).

## Differentiability at $x_0 = 230$ MeV



↪ The hit coordinate  $f(x)$  is differentiable in the beam energy  $x$  at  $x_0$ .

Derivatives in $\frac{\text{mm}}{\text{MeV}}$	• first layer	♦ second layer
Difference quotient	-0.082016	-0.130653

## Integrating Derivgrind into Geant4 with Client Requests

```
+ #include "/somepath/include/valgrind/derivgrind.h"
```

“Seeding” dot values of the input variable (beam energy):

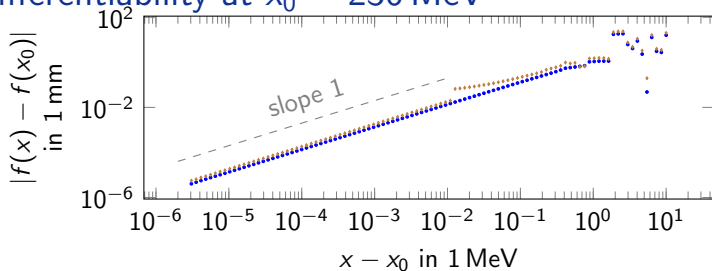
```
if (command == pIonCmd) {  
    pSourcePencilBeam->SetIonParameter(newValue);  
}  
if (command == pEnergyCmd) {  
    double energy = pEnergyCmd->GetNewDoubleValue(newValue);  
+   double one = 1.0;  
+   DG_SET_DOTVALUE(&energy, &one, sizeof(double));  
    pSourcePencilBeam->SetEnergy(energy);  
}
```

Obtaining dot values of output variables (hit position):

```
if (m_rootHitFlag) m_treeHit->Fill();  
+ float pos = *(float*)(m_treeHit->GetBranch("posX")->GetAddress());  
+ float pos_d;  
+ DG_GET_DOTVALUE(&pos, &pos_d, sizeof(float));  
+ std::cout << "pos_d=" << pos_d << "\n";
```

Running GATE under Derivgrind: `valgrind --tool=derivgrind Gate <args>`

## Differentiability at $x_0 = 230$ MeV



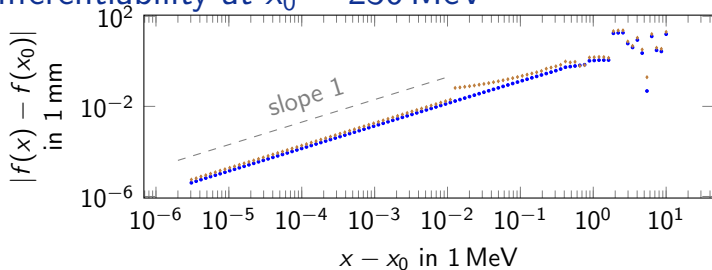
↪ The hit coordinate  $f(x)$  is differentiable in the beam energy  $x$  at  $x_0$ .

Derivatives in $\frac{\text{mm}}{\text{MeV}}$	• first layer	♦ second layer
Difference quotient	-0.082016	-0.130653
Derivgrind	-0.068512	-0.113841

## Bit-Tricks in Geant4

- Geant4 defines and uses a function `G4Log` adapted from the VDT math library.
- `G4Log` first performs a “range reduction”, scaling its argument by power of 2 to map it into  $[0.5, 1)$ .
- This is done by setting the exponent bits to `0b01111111110` via bitwise operations.
- Derivgrind’s normal dot value propagation does not recognize this as an arithmetic operation.
- **Fix:** Edit Geant4 source code, replacing body of `G4Log` by call to `log`.

## Differentiability at $x_0 = 230$ MeV



⇒ The hit coordinate  $f(x)$  is differentiable in the beam energy  $x$  at  $x_0$ .

Derivatives in $\frac{\text{mm}}{\text{MeV}}$	• first layer	♦ second layer
Difference quotient	-0.082016	-0.130653
Derivgrind	-0.068512	-0.113841
Derivgrind, G4Log replaced	-0.081339	-0.130524

⇒ After the fix, Derivgrind's derivative agrees well with difference quotients.

## Summary

- Derivgrind implements **forward-mode AD** and **operator-overloading-style reverse-mode AD** for compiled programs.
- Very limited access to primal source code needed, only in order to identify input/output variables (and for debugging in case of failure).
- In the case of unsupported **bit-tricks**, Derivgrind might miss or misunderstand real-arithmetic dependencies.
- Successfully tested on “big” software projects **Geant4**, **CPython** and **LibreOffice Calc**, with bit-trick issues observed in corner cases or easily fixable.

## Next Step

Debugging tools, applications, ...

## Contact Information

Max Aehle, [max.aehle@scicomp.uni-kl.de](mailto:max.aehle@scicomp.uni-kl.de),  
and Johannes Blühdorn, Max Sagebaum, Nicolas R. Gauger  
Chair for Scientific Computing  
University of Kaiserslautern-Landau (RPTU)

Code: <https://github.com/SciCompKL/derivgrind> ↗

Project E-Mail Address: [derivgrind@projects.rptu.de](mailto:derivgrind@projects.rptu.de)

Video (7 min) about Derivgrind+LibreOffice Calc:  
<https://t1p.de/tt4ne> ↗

