

Enabling Differentiable Rendering via AD of Parallel C++ STL Primitives in Clad

Midterm Evaluation

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Adding Concurrency and Inverse Rendering to Clad

Project Overview: To extend Clad's algorithmic differentiation capabilities to support multi-threading and path tracing applications.

Part 1: Concurrency (Forward-Mode AD)

- Propagating derivatives through reference wrappers.
- Adding differentiation support for `std::thread`.

Part 2: Inverse Rendering (Reverse-Mode AD)

- Resolving deep-cloning and loop-tracking compiler bugs.
- Building an initial soft-rasterization demo.
- Differentiating the traditional SmallPT path tracer.

Supporting Reference Wrappers (PR #1851)

Enabling `std::ref` and `std::cref` in Forward-Mode

Context

C++ threads copy arguments by default. Passing arguments by reference requires `std::ref`, but Clad needed to learn how to propagate derivatives through these wrapper objects.

Implementation

- Added custom pushforwards for `std::ref` and `std::cref` to generate derivative wrappers.
- Added support for `.get()` extraction to yield both primal and derivative values.

```
namespace std {  
    // Instructs Clad on how to push derivatives through std  
    template <class T>  
    inline clad::ValueAndPushforward<::std::reference_wrapper<T>  
        &::std::reference_wrapper<T>  
    ref_pushforward(T& t, T& d_t) noexcept {  
        // Returns a pair: primal reference and derivative ref  
        return {::std::ref(t), ::std::ref(d_t)};  
    }  
}
```

Adding Support for std::thread (PR #1891)

Forward-Mode AD for Concurrent Execution

Context

Using standard threading primitives previously caused Clad to generate invalid code or fail during the differentiation pass.

Implementation

- The AST visitor now recursively extracts the thread's callable payload and requests a derivative pushforward.
- Updates `utils::isCopyable` to detect deleted/private copy constructors, ensuring `std::move` semantics are used for non-copyable objects.

```
// 1. Primal C++ function dispatching work to a thread
double f_thread_ref(double x) {
    std::reference_wrapper<double> rx = std::ref(x);
    std::thread t(inc_ref, rx);
    t.join();
    return x;
}

// 2. Clad automatically generates the thread pushforward
int main() {
    auto d_thread = clad::differentiate(f_thread_ref, "x");

    // Evaluates both primal logic and derivative calculation
    printf("thread: %.4f\n", d_thread.execute(2.0));
    return 0;
}
```

Compiler Fixes for Reverse-Mode AD (PR #1876)

Resolving AST Deep-Cloning and Reverse-Forward Pass Issues

Context

This PR addresses two reverse-mode AD stability issues in Clad's AST cloning and reverse-forward pass handling, adding regression tests to prevent crashes in SmallPT-like code.

Changes

- Deep-clone argument expressions for CXXConstructExpr and CXXTemporaryObjectExpr in StmtClone to avoid sharing/mis-owning subexpressions.
- Add a loop guard in reverse-mode call handling to disable restore_tracker-dependent reverse-forward behavior when no tracker is available.

```
// lib/Differentiator/ReverseModeVisitor.cpp  
// Resolving tracker state bleed in reverse-mode loops  
  
if (isInsideLoop && usingRestoreTracker && !_RestoreTracker  
    calleeFnForwPassFD = nullptr;  
    usingRestoreTracker = false;  
}
```

Initial Inverse Rendering Demo (PR #1789)

Differentiable Geometry via Soft Rasterization

Context

Traditional path tracing relies on hard intersections, which create discontinuities that can return zero or NaN gradients in AD tools.

Approach

Introduced a "soft rasterization" model as a first step. By replacing hard boundaries with a Gaussian falloff, the math became smooth enough for Clad to calculate gradients and optimize scene parameters (camera, light, and objects).

```
// Differentiable approximation: soft-weight Gaussian falloff  
// Allows gradients to flow smoothly by avoiding hard geometry  
double ray_sphere_proximity(...) {  
    // ... distance math ...  
  
    // Smooth version: use sd^2 regardless of sign for gradient  
    double sd = dist - r;  
    return sd;  
}  
  
double soft_weight(double sd, double sharpness) {  
    // Gaussian falloff acts as the differentiable workaround  
    return exp(-sharpness * sd * sd);  
}
```

Differentiating Traditional SmallPT (PR #1877)

Applying Native AD to the Original Path Tracer

Objective

To demonstrate that Clad can differentiate the traditional SmallPT path tracer natively, without relying on the soft-rasterization workarounds.

Implementation

- Modularized the original SmallPT codebase.
- Ran reverse-mode AD directly through hard intersections, recursive bounces, and Russian roulette path termination.
- Added custom reverse-mode pullbacks for vector math (SmallPTDiffCommon.h) to optimize compiler workload, and validated gradients using Finite Difference (FD).

```
// SmallPTDiffRadiance.h
// Native differentiation through traditional path tracing
inline Vec diff_radiance(Ray r, int depth, unsigned short *Xi) {
    // ...
    if (!diff_intersect(r.o.x, r.o.y, r.o.z, r.d.x, r.d.y, r.d.z, Xi))
        return Vec(); // Hard miss

    // ...
    if (depth > 5 || !Xi) {
        // Clad successfully builds tapes through Russian roulette
        if (sample_erand48(Xi) < p) {
            Vec scaled_f = f * (1.0 / p);
            f = scaled_f;
        } else {
            return Vec();
        }
    }
}
```

Future Plan: 3DGS & Gather-Reduce

Transitioning from foundational capabilities to high-performance rendering optimization.

Phase 2: 3DGS Forward Redesign

Forking diff-gaussian-rasterization to strip out raw CUDA atomic loops.

- Implementing deterministic `thrust::sort_by_key`.
- Enforcing tile-based memory ownership via `thrust::reduce_by_key` to group pixel updates safely.

Phase 3: Clad Backward Synthesis

Triggering Clad's extended liveness analysis on the refactored renderer.

- Auto-generating the lock-free, zero-atomic backward pass from the Thrust abstractions.
- Validating gradients against finite differences and profiling memory throughput with Nsight Compute.