

# Cross-Language Component Testing: Performance and Interoperability Insights

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# Introduction

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- New developments in libraries/languages
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## Rewrites in software development

- Key component
- Thesis explores effective approaches

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## Objective

- Determine if the component works better in another language

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## Performance

How do performance attributes effect the application of interop APIs in performance dependent code?

# System design

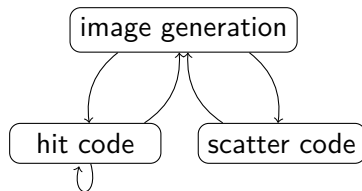


Figure: Ray tracer overview

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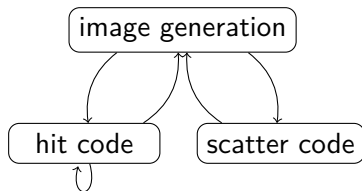


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- Complex enough system to demonstrate the idea in an applied setting.



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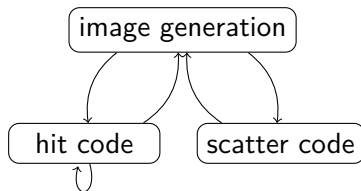


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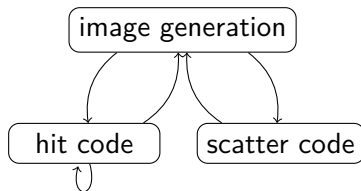


Figure: Ray tracer overview

- Complex enough system to demonstrate the idea in an applied setting.
- Distinct components that can be isolated for testing.
- Requires high performance to function effectively, making it ideal for assessing performance needs.

# Study Approach: Q&A Format

*How do we isolate the most performance impacting component?*

- Profile our application for runtime and memory.
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*How do we test the rewritten components?*

- Use the language specific benchmarking tools to test components in isolation.
- Use Julia's benchmarking tools to test components & overhead

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- Efficiency concern: If target languages perform poorly with hit functions, rewriting the entire ray tracer may not be feasible.
- Rewriting components: Trivial task due to similar syntax across languages.

## C++

```
bool hit(const aabb& box, const ray& r, const interval& ray_t);
```

## Python

```
def hit(bbox: aabb, r: ray, ray_t: interval[float]) -> bool:
```

## Julia

```
function hit!(bbox::aabb, r::ray, ray_t::interval)::Bool
```



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Languages supporting reflection reduce additional code, simplifying rewriting and testing.

# Testing setup

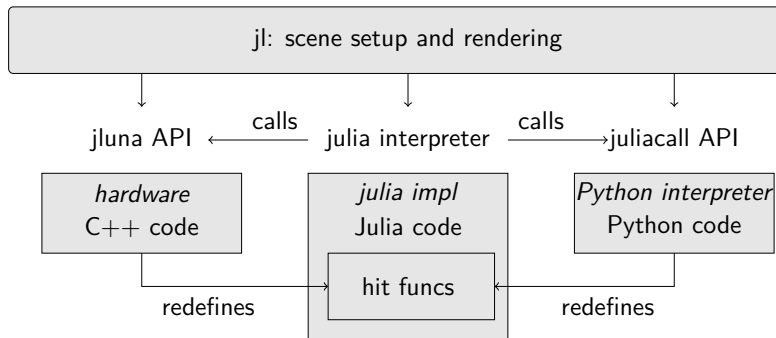


Figure: Testing setup for component isolation

# Mapping Objects to Statically Typed Languages

## Static Mapping

- User describes interface for types (e.g., class/struct).
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## Getters and Setters:

- Define functions `get` and `set`.
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**Note:** Similar to functional programming lens.

# Challenges with Getter/Setter Pairs

## Limitation

Problems occur when the setter has an abstract type.

### Problem:

- Julia uses strings to represent types.
- Need to convert these strings back to C++ types.

### Solution:

- Create a mechanism to map strings to types and compare them to find the right derived type

# Optimizing Setter Functions

Key issue with trivial approach

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*Preferable solution*

- Define the set of derived types once.
- Allow all setters for attributes of a class to access this information.

# Template Metafunctions for Object Properties

Template metafunctions are used to define object properties:

```
template <typename Ot, typename Ft, const char* name>
struct Property {
    static constexpr const char* get_name() { return name; }
    static std::function<Ft(Ot&)> getter;
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Example property declaration:

```
Usertype<bvh_node>::initialize_type(
    tl<
        Property<bvh_node, Hittable*, #left>,
        Property<bvh_node, Hittable*, #right>,
        Property<bvh_node, aabb, #bbox>
    >(),
    tl<Triangle, Sphere, bvh_node>()
);
```

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## Considerations

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### key challenge

Find some means of having the base type interface discern the correct derived type to instantiate.

# Benefits

This system offers

- Simplified object deduction process
- Ability to map abstract objects
- Utilization of modern C++ principles for futureproofing
- Increased maintainability and extensibility
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## Key Takeaway

It provides a more concise approach compared to traditional methods, enhancing ease of implementation and potential for future extensions.

# Insights - RQ2

## Observations

- Possible to implement a generic method for polymorphic object mapping.
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## Main takeaways

- We can map to static languages with minimal additional boilerplate
- Mostly straightforward testing and integration across languages

# Baseline Performance analysis

## Timing method

- Call to `bvh_hit` from Julia
- Execution time in target languages (C++ and Python)

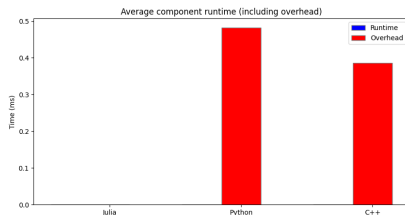
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## Results

- Julia
  - Fastest average component execution time since no overhead
- Python and C++:
  - Similar average component execution times
  - Execution time primarily attributed to overhead





# Isolated Performance - RQ3

## Results

- C++ outperforms Julia and Python in component runtime.
- Substantial overhead from Julia API (dynamic dispatch, type inference).
- Julia performs well with type stability, hindered by cross-language calls.

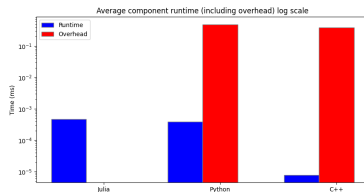
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## Conclusion

- High-performance interop between different languages is challenging.
- Effective on a smaller scale, but less viable for very different languages.



# Results and Takeaways

- Different language APIs have varied integration approaches.
- Adapting APIs to specific use cases is generally manageable depending on the language
- Metaprogramming is effective for implementing generic, extensible interop libraries.
- Performance overhead can depends alot on language compatability and paradigms
- APIs work well for benchmarking and testing components in isolation
- Careful consideration needed for using these APIs in performance-dependent production code.