# DiffOptics: Imaging Simulation with Differentiable Ray Tracing

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# Introduction

Optics and imaging simulator development began within context of the Diagnostic Imaging System SLAC is developing

Goals:

- Test different setups, lenses, NA, ...
- Understand impact of noise, QE, ...
- Understand geometric aberration, Depth of Field, PSF, ...
- Estimate wave function phase fits and system capabilities
- Develop new imaging system ideas

Beyond typical optics simulation, want to have simulator gradients:

- Want to use modern ML and rendering methods, and hardware (GPU)
- Want to be able to easily optimize and calibrate optics systems
- *Long term*: Differentiable simulation pipeline optimizable with gradients, and usable within analysis chain?



### Overview

#### **Differentiable-Optics (diffoptics):** Simulator of optical system and image generation

- Geometric Optics Approach: Ray tracing
- Diffraction effects from Point Spread Function

Differentiable code enables new optimization and inference schemes

#### Inference

- Simulator can be used at inference time, i.e. solve A(x)=b
- Can be easily integrated into an ML pipeline



### Software and differentiability $\rightarrow$ Differentiable Programming 4

#### Python with PyTorch (deep learning framework ) backend

- Vectorized code, works on CPU and GPU
- Automatic differentiation framework

#### Simulations are fully differentiable

- When we write a function: f(x) {...}; automatically get function to evaluate the derivative: df(x) {...};
- Can use gradients for e.g. calibration, reconstruction, optimization...
- Seamlessly used in Machine Learning pipelines

Software design based on graphics / rendering engines

### **Optics elements**

Optical elements organized in a "scene": Positions and orientations can be specified

Optical Elements implemented:

- Ideal (thin) lens and thick lens (in progress)
- Mirrors
- Windows
- Sensor
  - -Can specify resolution, quantum efficiency, noise
  - -Ongoing work to add electronics noise model
- Bounding box

Every optical element has a get\_ray\_intersection method to determine where rays interact with elements



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# Atom Clouds and Sampling Tools

Atom cloud shape can be specified

- "Standard" wave function implemented
- Can specify density, position, imaging time, etc.
- Can specify a method to sample rays from cloud

For arbitrary cloud densities, no direct method to sample rays

- Rejection sampling in (specifiable) bounded area
- Importance sampling in (specifiable) bounded area







#### Forward ray tracing (Object to Sensor)

• Sample ray position and direction, trace forward through system

#### Reverse Ray tracing (Sensor to Object)

- Position and angle sampling from each pixel
- Integration of cloud density along rays in (specifiable) bounded area

Vectorization and GPUs allow us to greatly speed up this process





## **Point Spread Functions**

#### PSF implemented as image post-processing step

- No diffraction / interference in ray tracing
- Impact implemented through convolution
- Position and depth dependent PSF can be specified



#### PSF must be measured / simulated elsewhere

- No first principles calculation (could be included if desired)
- Many lenses we want to buy do not have open source models in any case

#### Image generation + Fit



### **Probing Depth of Field**



#### Application to MAGIS Diagnostic Imaging System Design: Testing different setups



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### Fit directly with Differentiable Simulator



Can also fit phase directly through simulator

Cloud( $\phi$ )  $\rightarrow$  Simulated image  $X_{\phi} \rightarrow$  Compare to data Y (e.g. with Likelihood)

$$\phi^* = \arg\min_{\phi} L(Y, X_{\phi})$$

Simulated images are differentiable w.r.t. optical element parameters (size, shape, position, tilt...)

Can define optimization objective for specific goals

- Design Optimization
- Simulator Calibration
- Optimal Experiment Design

Can do optimization with gradient descent!



on sensor during optimization



# **Getting Started**

"Getting Started" tutorial provided in Jupyter Notebook

Singularity and Docker containers with all needed dependencies are available

Can run on CPU and GPU

Code currently in private GitHub repo

- Happy to give access
- Still plenty to do  $\rightarrow$  come contribute
- Aim to make it public soon!

```
import diffoptics as optics
sensor = optics.Sensor(position=(-f * (1 + m), 0, 0), poisson_noise_mean=2)
scene = optics.Scene(sensor)
lens = optics.PerfectLens(f=f, m=m)
scene.add_object(lens, False)
atom_cloud = optics.AtomCloud(position=(cloud_x_pos, 0., 0.))
atom_cloud = optics.LightSourceFromDistribution(atom_cloud)
for batch in range(20): # Trace 1B rays with minibatches of 50M rays
rays = atom_cloud.sample_rays(50e6, device='cuda)
photon_mapping(rays, scene)
img = sensor.readout()
```

```
# Study the impact of DoF
import numpy as np
import diffoptics as optics
sensor = optics.Sensor(position=(-f * (1 + m), 0, 0), poisson_noise_mean=2)
scene = optics.Scene(sensor)
lens = optics.PerfectLens(f=f, m=m)
scene.add_object(lens, False)
for delta_x in np.linspace(-1, 1, num=100):
    atom_cloud = optics.AtomCloud(position=(cloud_x_pos + delta_x, 0., 0.))
    atom_cloud = optics.LightSourceFromDistribution(atom_cloud)
    for batch in range(20): # Trace 1B rays with minibatches of 50M rays
    rays = atom_cloud.sample_rays(50e6, device='cuda)
    photon_mapping(rays, scene)
    img = sensor.readout()
```

# Conclusion

Active development of fully differentiable ray tracing simulator

- Novelty: merging differentiable ray tracing and optics
- Not MAGIS specific, useful to other experiments and applications

Broad applicability from physics imaging to experiment design optimization

Near-term plans to expand

- PSF recently merged into main code
- Implementation of optical elements to model compound systems with thick lenses
- Transmission / reflectance
- Could include optical path length for estimating diffraction effects