MEP Literature review

Designing free-form optics for multiple source illumination using differentiable ray-tracing and neural networks

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Caustic design

Caustic design

Def: A part of non-imaging optics concerned with the geometrical design of optical components to transform the incident illumination from light sources into a desired target distribution.¹



Figure: An example of caustic design: approximating the painting *A Sunday Afternoon on the Island of La Grande Jatte* by Georges Seurat.²

¹J. Meyron, Q. Mérigot, and B. Thibert (2018). "Light in Power: A general and parameter-free algorithm for caustic design". In: ACM Transactions on Graphics 37 (6). ISSN: 15577368. DOI: 10.1145/3272127.3275056

²M. Nimier-David, D. Vicini, T. Zeltner, and W. Jakob (Dec. 2019b). "Mitsuba 2: A Retargetable Forward and Inverse Renderer". In: *Transactions on Graphics (Proceedings of SIGGRAPH Asia)* 38.6. DOI: 10.1146/3355089.33556498

Previous work

- Lucas' work on Caustic design with MLP + B-spline surface + wave optics. 3
- Differentiable rendering in Mitsuba 2.4

³L. H. Crijns (2021). PINN inspired Freeform Design. Delft University of Technology

⁴M. Nimier-David, D. Vicini, T. Zeltner, and W. Jakob (2019a). "Mitsuba 2". In: ACM Transactions on Graphics 38 (6), pp. 1–17. ISSN: 0730-0301. DOI: 10.1145/3355089.3356498

Optimization pipeline



Scientific Machine Learning



Scientific Machine Learning: PINNs



Figure: PINN schematic.⁵

⁵X. Meng, Z. Li, D. Zhang, and G. E. Karniadakis (Sept. 2019). "PPINN: Parareal Physics-Informed Neural Network for time-dependent PDEs". In: DOI: 10.1016/j.cma.2020.113250

(Scientific Machine Learning): Inverse problem

- Forward problem: Lens \Rightarrow caustic
- Inverse problem: Lens ← caustic





 \Rightarrow Analysis-by-synthesis

Spline surfaces



Spline surfaces: Plots



Spline surfaces: Definitions

B-spline surface:

$$\mathbf{S}(u,v) \coloneqq \sum_{i=0}^{n_1} \sum_{j=0}^{n_2} N_{i,p}(u) N_{j,q}(v) \mathbf{P}_{i,j}$$
(1)

NURBS surface:

$$\mathbf{S}(u,v) \coloneqq \sum_{i=0}^{n_1} \sum_{j=0}^{n_2} \frac{w_{i,j} N_{i,p}(u) N_{j,q}(v)}{\sum_{i'=0}^n \sum_{j'=0}^m w_{i',j'} N_{i',p}(u) N_{j',q}(v)} \mathbf{P}_{i,j}$$
(2)



Standard work on NURBS: NURBS book.⁶

⁶L. Piegl and W. Tiller (1997). *The NURBS Book*. Springer Berlin Heidelberg. ISBN: 978-3-540-61545-3. DOI: 10.1007/978-3-642-59223-2

Ray-tracing



Ray-tracing: Overview

Important scene components

- Light source(s)
- Lens mesh
- BSDF: Bi-directional scattering distribution function
- Receiver screen (diffuser)
- Camera



Figure: Schematic setup of a ray-tracer I wrote



Ray-tracing: BSDF



⁷M. Pharr, W. Jakob, and G. Humphreys (2017). *Physically Based Rendering*. DOI: 10.1016/b978-0-12-800645-0.50001-4

Ray-tracing: Optics basics⁸



(a) **Reflection** angles: $\theta_i = \theta_r$

vectors: $\hat{\mathbf{t}} = \hat{\mathbf{s}} - 2\langle \hat{\mathbf{s}}, \hat{\mathbf{n}} \rangle \hat{\mathbf{n}}$



(b) Refraction angles: $n_i \sin \theta_i = n_t \sin \theta_t$ vectors: $\hat{\mathbf{t}} = \frac{n_i}{n_t} \hat{\mathbf{s}} - \left(\frac{n_i}{n_t} \langle \hat{\mathbf{s}}, \hat{\mathbf{n}} \rangle + \sqrt{1 - \left(\frac{n_i}{n_t}\right)^2 (1 - \langle \hat{\mathbf{s}}, \hat{\mathbf{n}} \rangle^2)}\right) \hat{\mathbf{n}}$

⁸L. B. Romijn (2021). Generated Jacobian Equations in Freeform Optical Design. ISBN: 9789464166941

Optimization



Optimization: Landscape

Loss = $\|$ Current illumination – Target illumination $\|^2$



Figure: Example of a design landscape: loss (z) as a function of network parameters (x, y).⁹

Optimizers: Adam¹⁰, L-BFGS¹¹

⁹J. Berner, P. Grohs, G. Kutyniok, and P. Petersen (May 2021). "The Modern Mathematics of Deep Learning".

¹⁰D. P. Kingma and J. Ba (Dec. 2014). "Adam: A Method for Stochastic Optimization". In

 $^{11}\text{D.}$ C. Liu and J. Nocedal (1989). On the limited memory BFGS method for large scale optimization. Pp. 503–528

Optimization: Automatic Differentiation



(a) Gradient of the surface with respect to the z-coordinate of a control point.



Forward-rendering result

ŤUDelft

Differentiable-rendering result

(b) Differentiable rendering example: derivative w.r.t. sun location. Source: https://diff-render.org/.

Figure: AD examples

Optimization: Inverse rendering



Figure: Inverse rendering without neural network.¹²

¹²M. Nimier-David, D. Vicini, T. Zeltner, and W. Jakob (Dec. 2019b). "Mitsuba 2: A Retargetable Forward and Inverse Renderer". In: *Transactions on Graphics (Proceedings of SIGGRAPH Asia)* 38.6. DOI: 10.1145/3355089.3356498

Research: Considerations

Hyper-parameters

- Network: width, depth, connectivity
- Sources: type and amount
- Spline surface(s): control net size, degrees

(Un)free parameters

- Control point z coordinates
- Control point *x*, *y* coordinates
- Control point weights
- Source location(s)

Research: Starting model

Starting model

- One simple (zero-etendue) fixed light source: point or plane
- Specularly scattering lens mesh with one B-spline surface, n = 1.5
- A simple MLP with a convolutional layer reflecting the B-spline basis function support overlap, only control point *z*-coordinates are output
- A loss function consisting of a RMS-difference of the current and target render
- The L-BFGS optimizer



Figure: Example lens mesh.

Research: Goals

Research goals

- See whether the lens that produces a certain caustic can be found back
- 2 See what the influence of training with weights is on convergence
- See whether a lens for desired smooth caustics, *e.g.* for street lighting, can be found (and for which surface hyper-parameters)