

3D Gaussian Splatting for Real-Time Radiance Field Rendering

Algorithm

Input – set of images of static scene

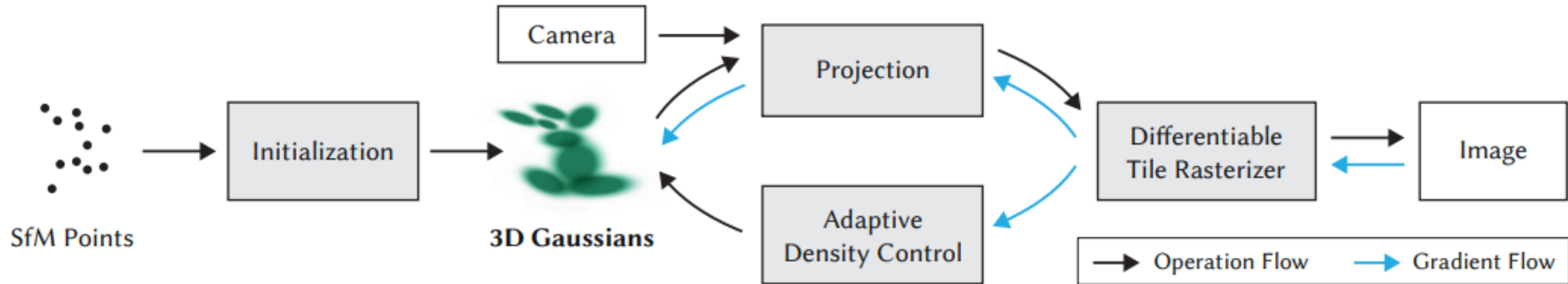
1. Get sparse point cloud by SfM
2. Create a set of 3D Gaussians
3. Optimize parameters of Gaussians with gradient descent
4. Render an image

Sparse point cloud



Pipeline

3D Gaussian Splatting for Real-Time Radiance Field Rendering • 1:5



Parameters of a Gaussian

- Position (mean)

$$G(x) = e^{-\frac{1}{2}(x)^T \Sigma^{-1}(x)}$$

- Covariance matrix
(3x3)

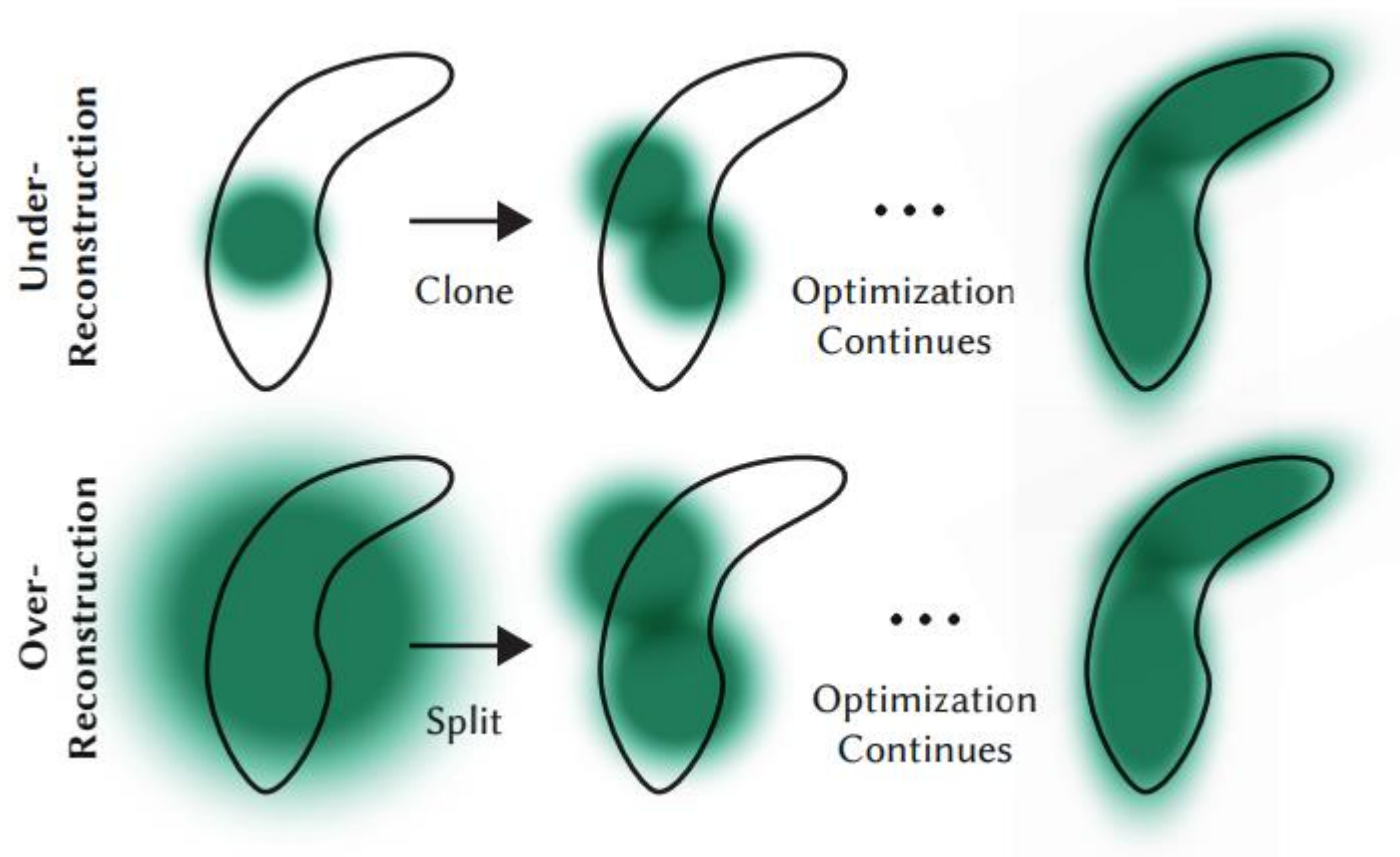
$$\Sigma = RSS^T R^T - \text{should be positive semi-definite}$$

R – rotation matrix

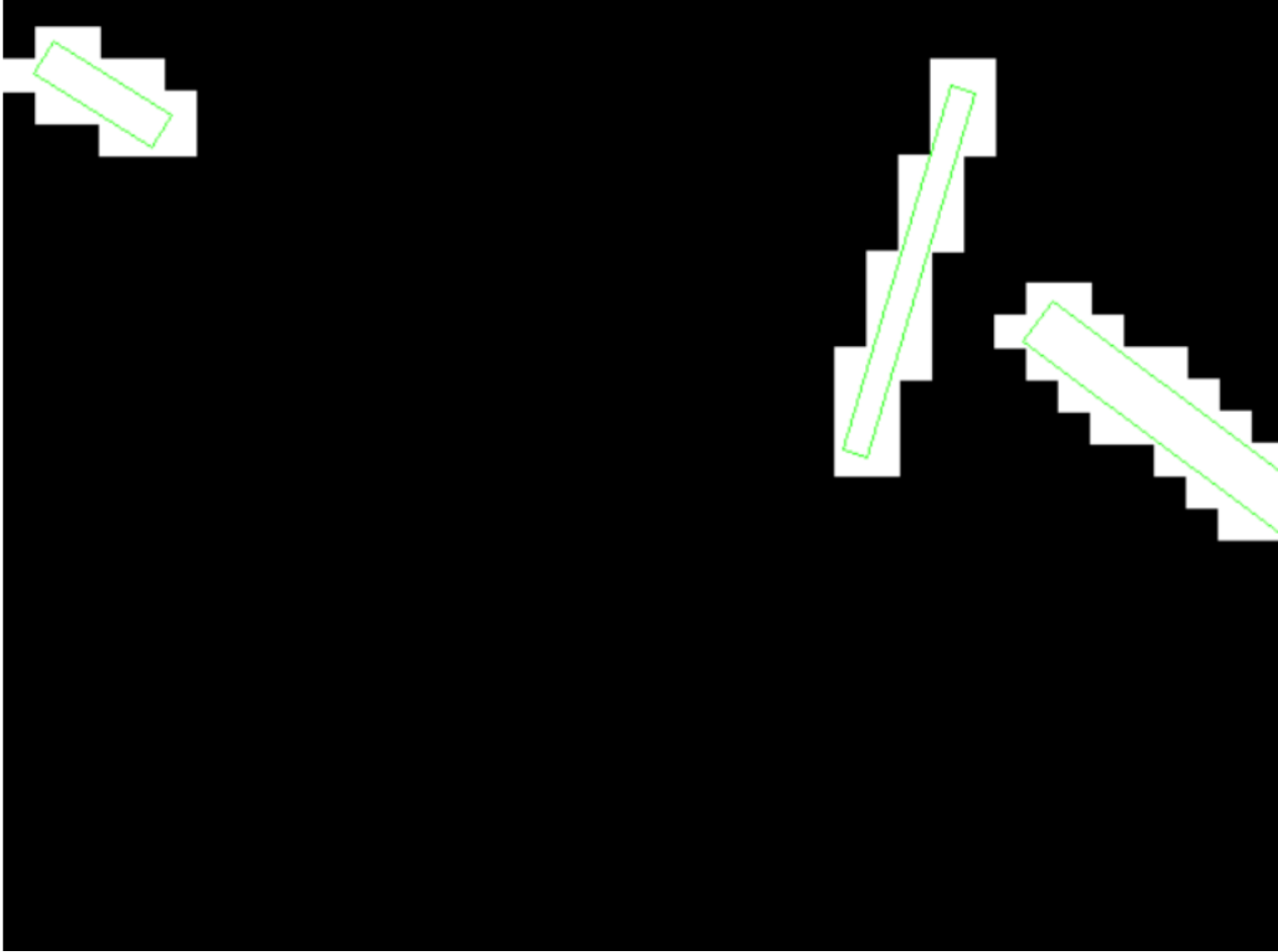
S – scale matrix

- Color (RGB)
- Opacity ([0, 1])

Gaussian control



Rasterizer



$$C = \sum_{i \in \mathcal{N}} c_i \alpha_i \prod_{j=1}^{i-1} (1 - \alpha_j)$$

α_i - opacity of i-th point

c_i - color of i-th point

Algorithm 1 Optimization and Densification w, h : width and height of the training images

```

 $M \leftarrow$  SfM Points ▷ Positions
 $S, C, A \leftarrow$  InitAttributes() ▷ Covariances, Colors, Opacities
 $i \leftarrow 0$  ▷ Iteration Count
while not converged do
   $V, \hat{I} \leftarrow$  SampleTrainingView() ▷ Camera  $V$  and Image
   $I \leftarrow$  Rasterize( $M, S, C, A, V$ ) ▷ Alg. 2
   $L \leftarrow$  Loss( $I, \hat{I}$ ) ▷ Loss
   $M, S, C, A \leftarrow$  Adam( $\nabla L$ ) ▷ Backprop & Step
  if IsRefinementIteration( $i$ ) then
    for all Gaussians  $(\mu, \Sigma, c, \alpha)$  in  $(M, S, C, A)$  do
      if  $\alpha < \epsilon$  or IsTooLarge( $\mu, \Sigma$ ) then ▷ Pruning
        RemoveGaussian()
      end if
      if  $\nabla_p L > \tau_p$  then ▷ Densification
        if  $\|S\| > \tau_S$  then ▷ Over-reconstruction
          SplitGaussian( $\mu, \Sigma, c, \alpha$ )
        else ▷ Under-reconstruction
          CloneGaussian( $\mu, \Sigma, c, \alpha$ )
        end if
      end if
    end for
  end if
   $i \leftarrow i + 1$ 
end while

```

Algorithm 2 GPU software rasterization of 3D Gaussians w, h : width and height of the image to rasterize M, S : Gaussian means and covariances in world space C, A : Gaussian colors and opacities V : view configuration of current camera

```

function RASTERIZE( $w, h, M, S, C, A, V$ )
  CullGaussian( $p, V$ ) ▷ Frustum Culling
   $M', S' \leftarrow$  ScreenspaceGaussians( $M, S, V$ ) ▷ Transform
   $T \leftarrow$  CreateTiles( $w, h$ )
   $L, K \leftarrow$  DuplicateWithKeys( $M', T$ ) ▷ Indices and Keys
  SortByKeys( $K, L$ ) ▷ Globally Sort
   $R \leftarrow$  IdentifyTileRanges( $T, K$ )
   $I \leftarrow 0$  ▷ Init Canvas
  for all Tiles  $t$  in  $I$  do
    for all Pixels  $i$  in  $t$  do
       $r \leftarrow$  GetTileRange( $R, t$ )
       $I[i] \leftarrow$  BlendInOrder( $i, L, r, K, M', S', C, A$ )
    end for
  end for
  return  $I$ 
end function

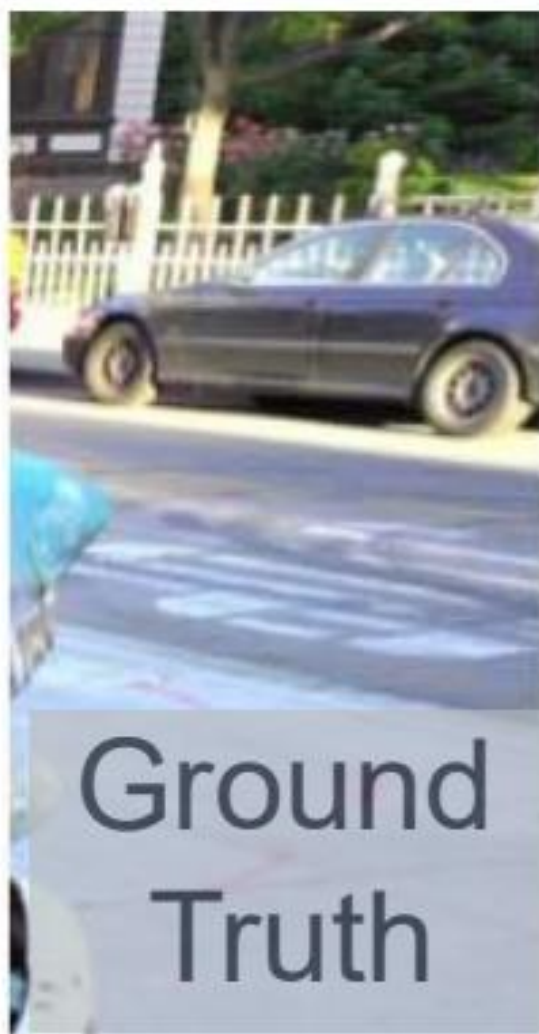
```

Metrics

Dataset	Mip-NeRF360						Tanks&Temples						Deep Blending					
Method Metric	<i>SSIM</i> [↑]	<i>PSNR</i> [↑]	<i>LPIPS</i> [↓]	Train	FPS	Mem	<i>SSIM</i> [↑]	<i>PSNR</i> [↑]	<i>LPIPS</i> [↓]	Train	FPS	Mem	<i>SSIM</i> [↑]	<i>PSNR</i> [↑]	<i>LPIPS</i> [↓]	Train	FPS	Mem
Plenoxels	0.626	23.08	0.463	25m49s	6.79	2.1GB	0.719	21.08	0.379	25m5s	13.0	2.3GB	0.795	23.06	0.510	27m49s	11.2	2.7GB
INGP-Base	0.671	25.30	0.371	5m37s	11.7	13MB	0.723	21.72	0.330	5m26s	17.1	13MB	0.797	23.62	0.423	6m31s	3.26	13MB
INGP-Big	0.699	25.59	0.331	7m30s	9.43	48MB	0.745	21.92	0.305	6m59s	14.4	48MB	0.817	24.96	0.390	8m	2.79	48MB
M-NeRF360	0.792 [†]	27.69 [†]	0.237 [†]	48h	0.06	8.6MB	0.759	22.22	0.257	48h	0.14	8.6MB	0.901	29.40	0.245	48h	0.09	8.6MB
Ours-7K	0.770	25.60	0.279	6m25s	160	523MB	0.767	21.20	0.280	6m55s	197	270MB	0.875	27.78	0.317	4m35s	172	386MB
Ours-30K	0.815	27.21	0.214	41m33s	134	734MB	0.841	23.14	0.183	26m54s	154	411MB	0.903	29.41	0.243	36m2s	137	676MB

	Truck-5K	Garden-5K	Bicycle-5K	Truck-30K	Garden-30K	Bicycle-30K	Average-5K	Average-30K
Limited-BW	14.66	22.07	20.77	13.84	22.88	20.87	19.16	19.19
Random Init	16.75	20.90	19.86	18.02	22.19	21.05	19.17	20.42
No-Split	18.31	23.98	22.21	20.59	26.11	25.02	21.50	23.90
No-SH	22.36	25.22	22.88	24.39	26.59	25.08	23.48	25.35
No-Clone	22.29	25.61	22.15	24.82	27.47	25.46	23.35	25.91
Isotropic	22.40	25.49	22.81	23.89	27.00	24.81	23.56	25.23
Full	22.71	25.82	23.18	24.81	27.70	25.65	23.90	26.05





Problems

- Artifacts in regions where the scene is not well observed
- Memory consumption

References

- <https://arxiv.org/abs/2308.04079v1> - main article
- <https://github.com/graphdeco-inria/gaussian-splatting> - implementation
- https://github.com/joeyan/gaussian_splatting/blob/main/MATH.md - math of gaussian splatting
- <https://huggingface.co/blog/gaussian-splatting> - hugging-face review
- <https://habr.com/ru/articles/768590/> - habr review