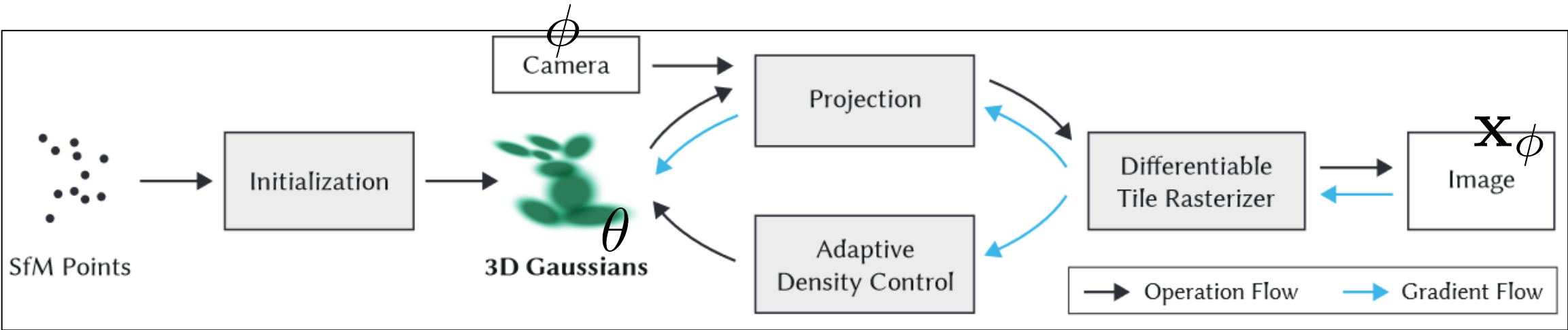
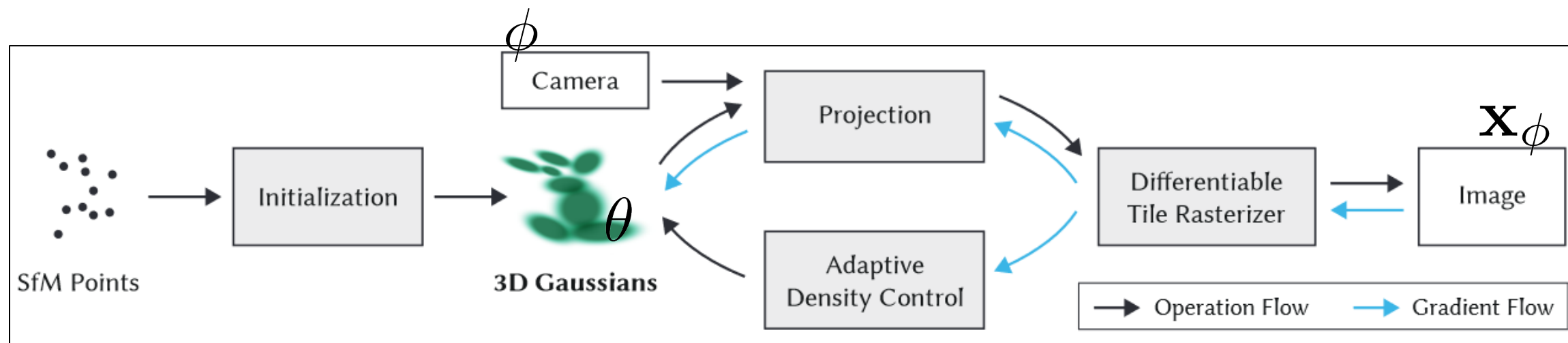




Dataset Method Metric	Mip-NeRF360			Train	FPS	Mem
	<i>SSIM</i> [↑]	<i>PSNR</i> [↑]	<i>LPIPS</i> [↓]			
Plenoxels	0.626	23.08	0.463	25m49s	6.79	2.1GB
INGP-Base	0.671	25.30	0.371	5m37s	11.7	13MB
INGP-Big	0.699	25.59	0.331	7m30s	9.43	48MB
M-NeRF360	0.792 [†]	27.69 [†]	0.237 [†]	48h	0.06	8.6MB
Ours-7K	0.770	25.60	0.279	6m25s	160	523MB
Ours-30K	0.815	27.21	0.214	41m33s	134	734MB

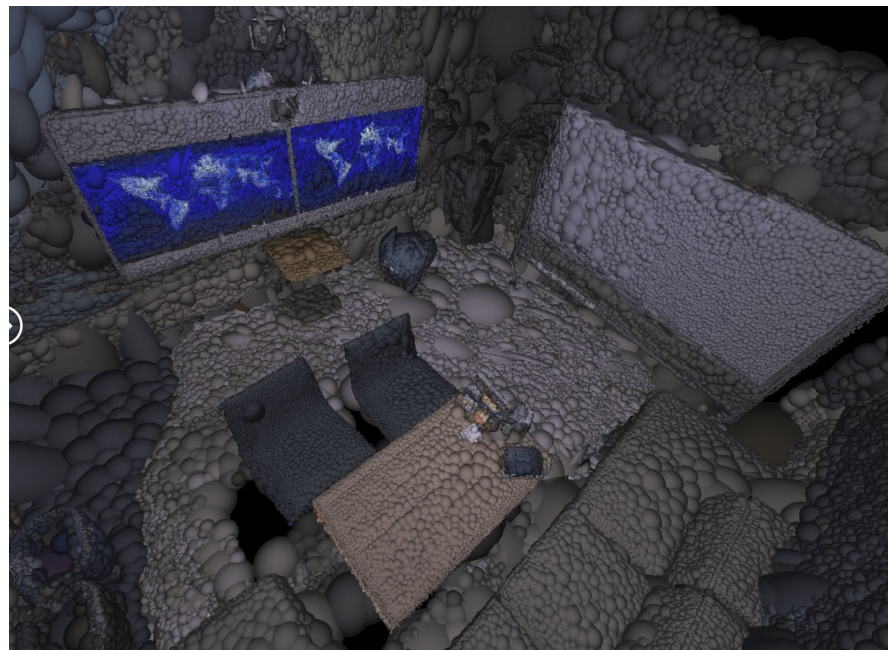


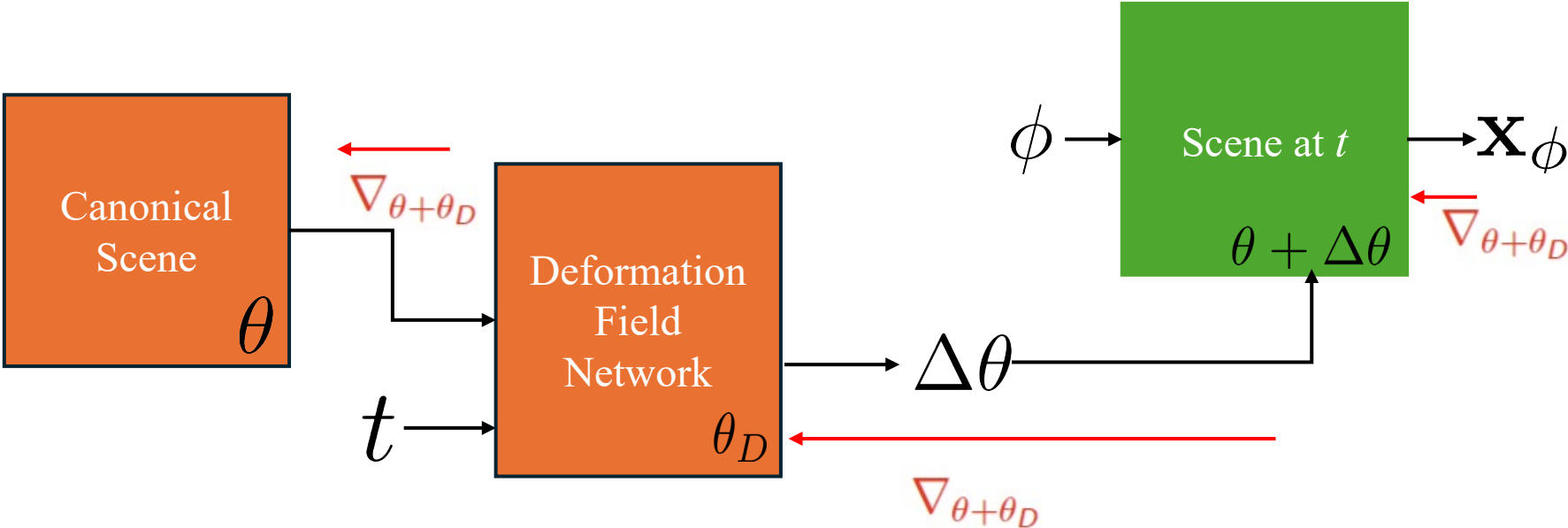


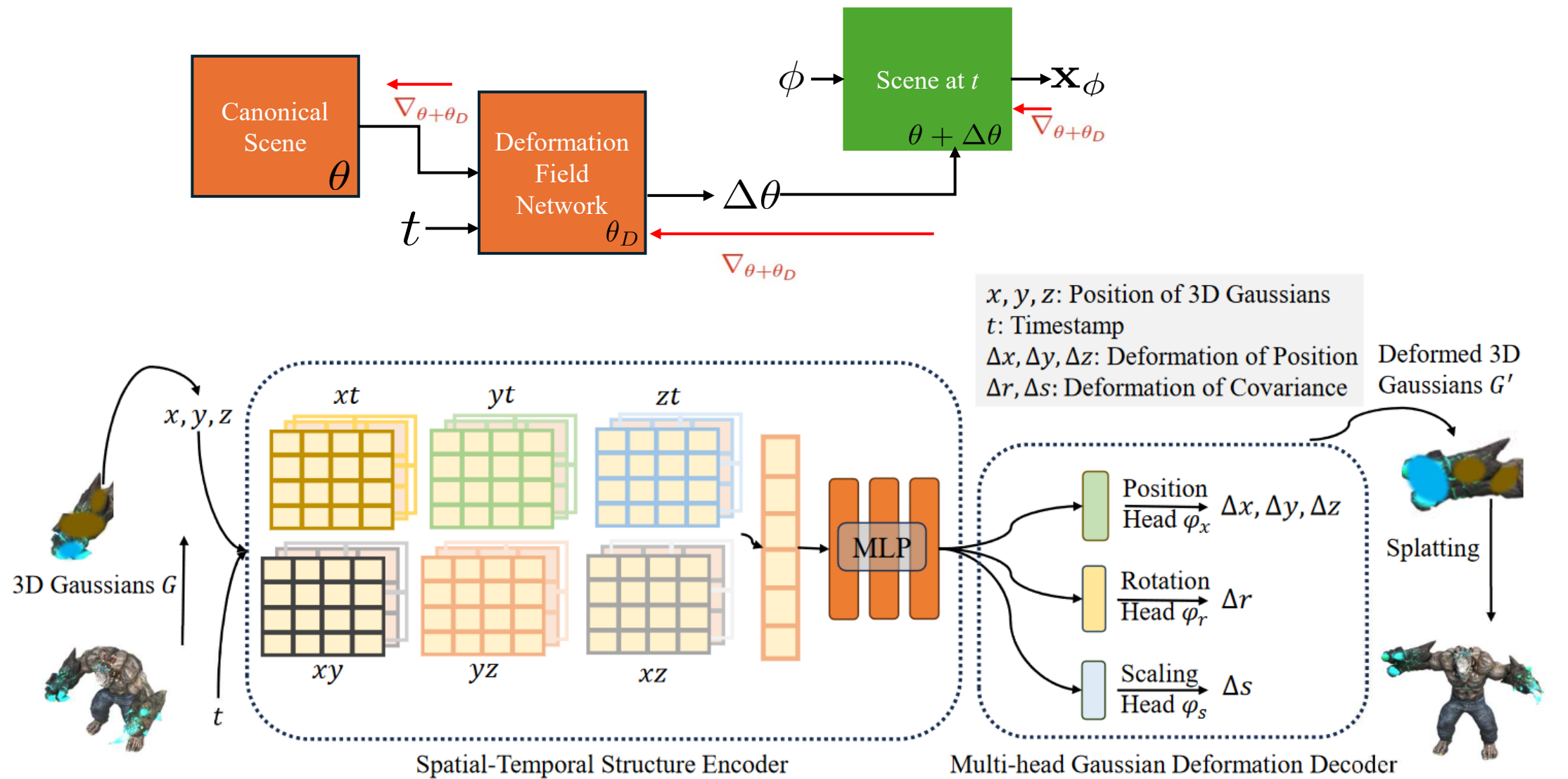
θ : a set of $\left[\underbrace{\mu_i}_{\text{position} \in \mathcal{R}^3}, \underbrace{\sigma_i}_{\text{opacity} \in \mathcal{R}^1}, \underbrace{\Sigma_i = (r_i, s_i)}_{\text{covariance} \in \mathcal{R}^7}, \underbrace{\mathbf{c}_i}_{\text{spherical harmonics} \in \mathcal{R}^{(k+1)^2+1}} \right]$

Projection: $\mu_i^{2D} = (K((W\mu_i)/(W\mu_i)_z))_{1:2}$,
 $\Sigma_i^{2D} = (JW\Sigma_i W^T J^T)_{1:2,1:2}$, approximation of perspective transform W, K

Rasterization: $\mathbf{I} = \sum_{i \in \mathcal{N}} \mathbf{c}_i \alpha_i^{2D} \prod_{j=1}^{i-1} (1 - \alpha_j^{2D})$ point-based blending of ordered points v.s. sampling rays in NeRF

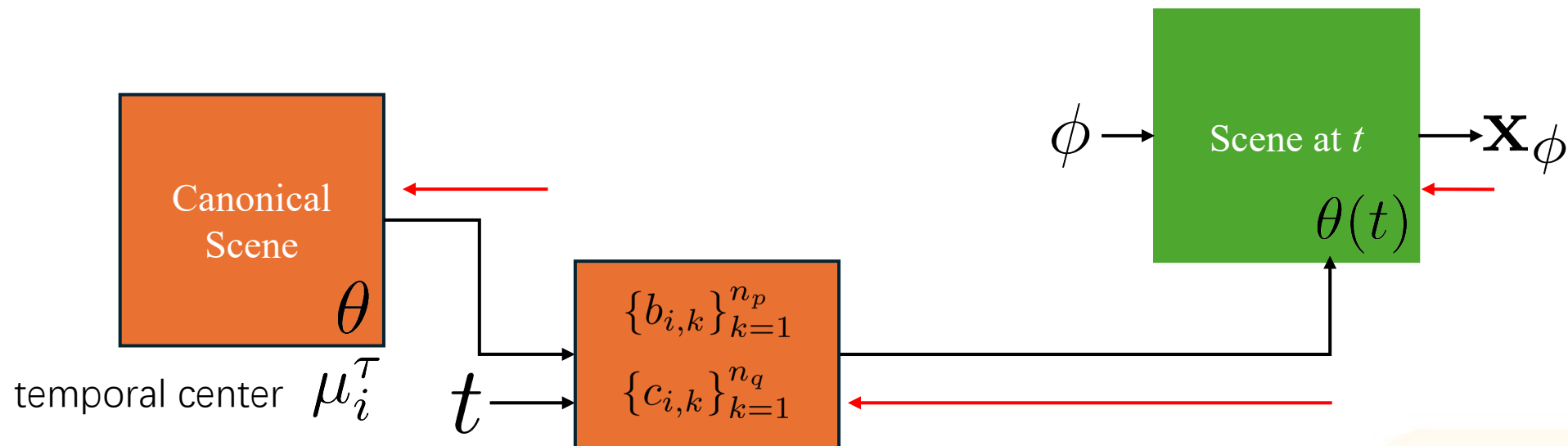






Hexplane Parameter: $R(i, j) \in R^{h \times l \times N_i \times N_j}$

Separate MLP for position, rotation, scaling



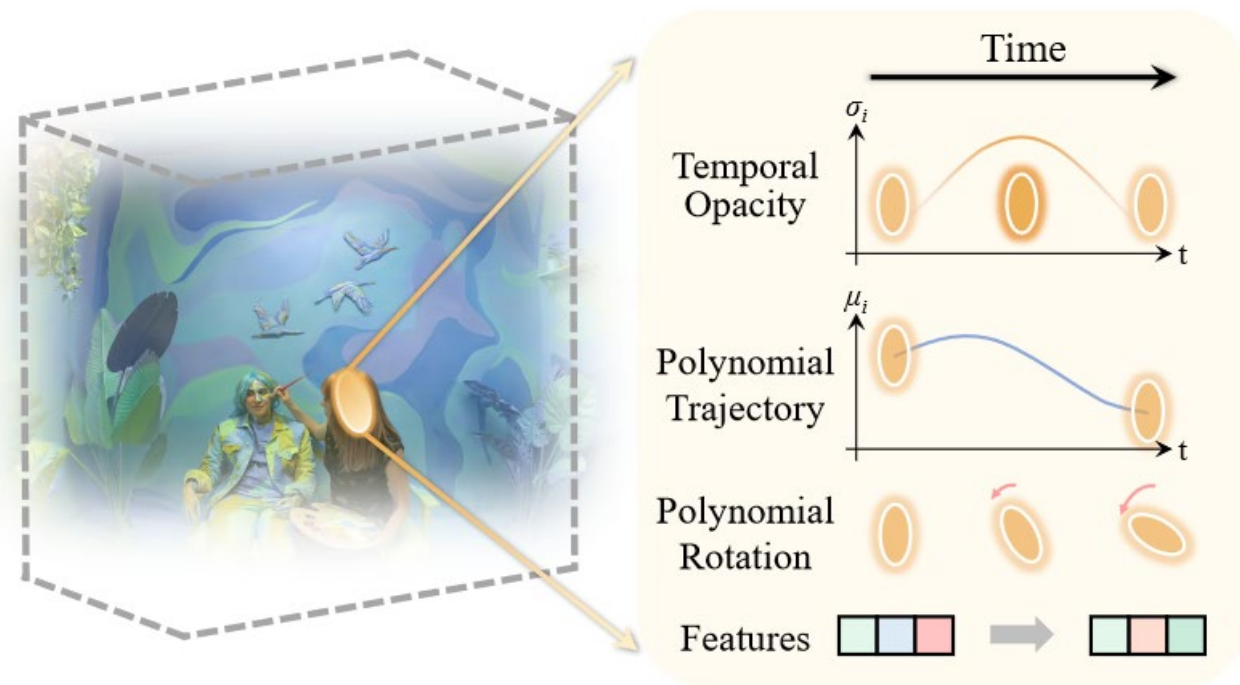
Polynomial Interpolation around Temporal Center

Time-dependent position

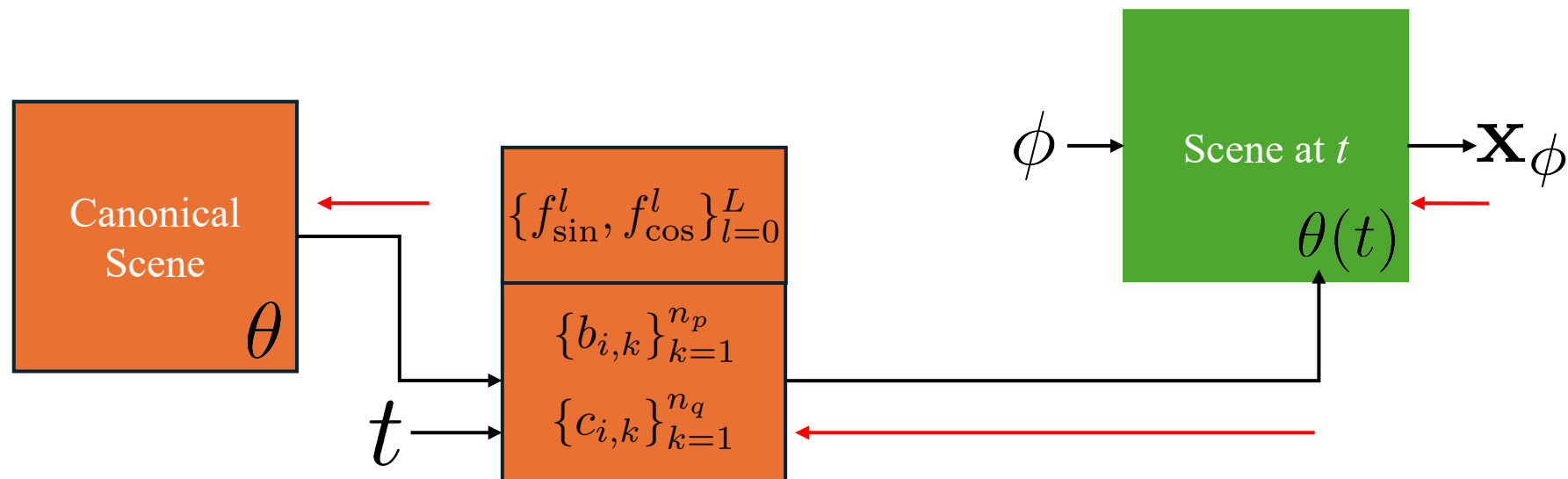
$$\mu_i(t) = \sum_{k=0}^{n_p} b_{i,k} (t - \mu_i^\tau)^k,$$

Time-dependent rotation

$$q_i(t) = \sum_{k=0}^{n_q} c_{i,k} (t - \mu_i^\tau)^k,$$



(a) Spacetime Gaussians



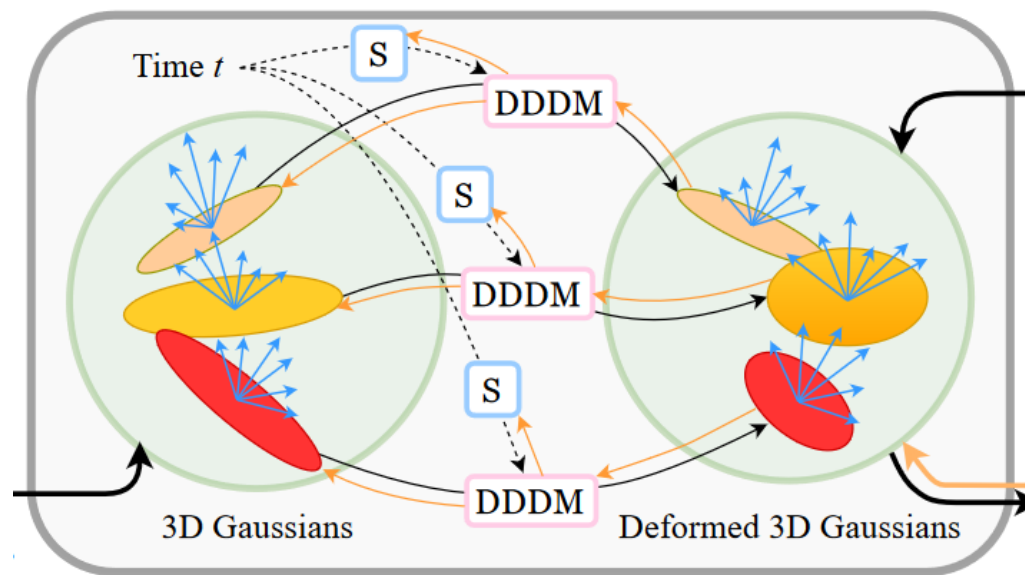
Dual-Domain Deformation Model

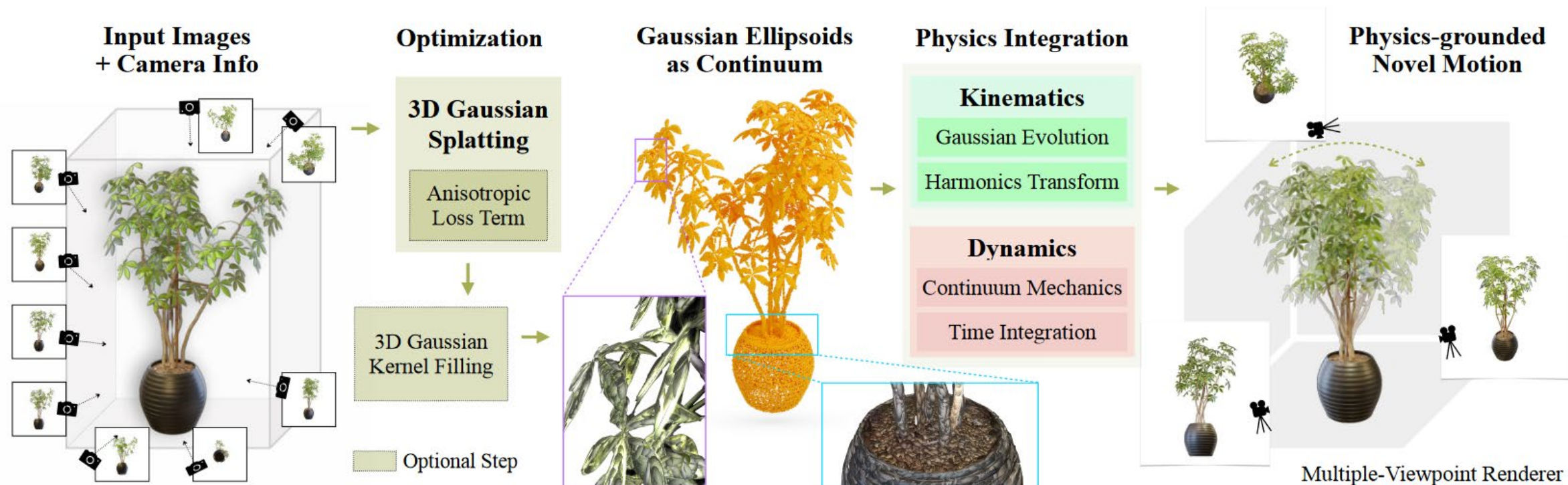
polynomial fitting in **time domain**

$$P_N(t) = \sum_{n=0}^N a_n t^n,$$

Fourier series fitting in **frequency domain**

$$F_L(t) = \sum_{l=1}^L (f_{\sin}^l \cos(lt) + f_{\cos}^l \sin(lt))$$





Continuum Mechanics

Conservation of mass

$$\int_{B_\epsilon^t} \rho(\mathbf{x}, t) \equiv \int_{B_\epsilon^0} \rho(\phi^{-1}(\mathbf{x}, t), 0),$$

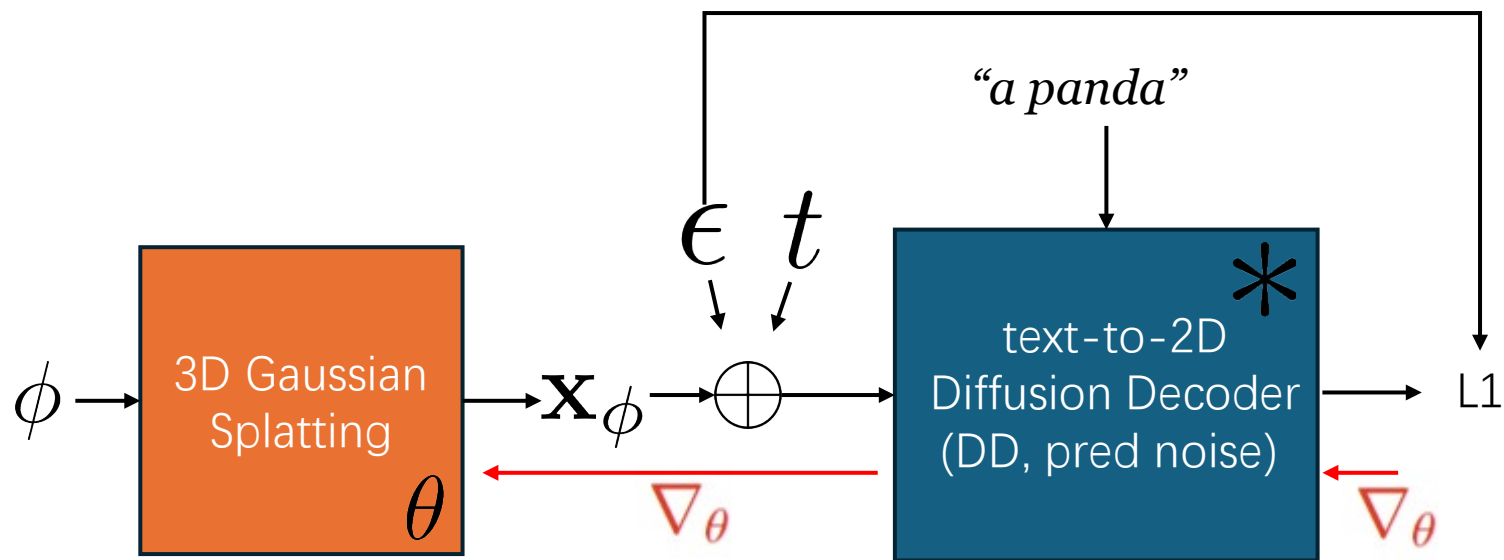
conservation of momentum

$$\rho(\mathbf{x}, t) \dot{\mathbf{v}}(\mathbf{x}, t) = \nabla \cdot \boldsymbol{\sigma}(\mathbf{x}, t) + \mathbf{f}^{\text{ext}},$$

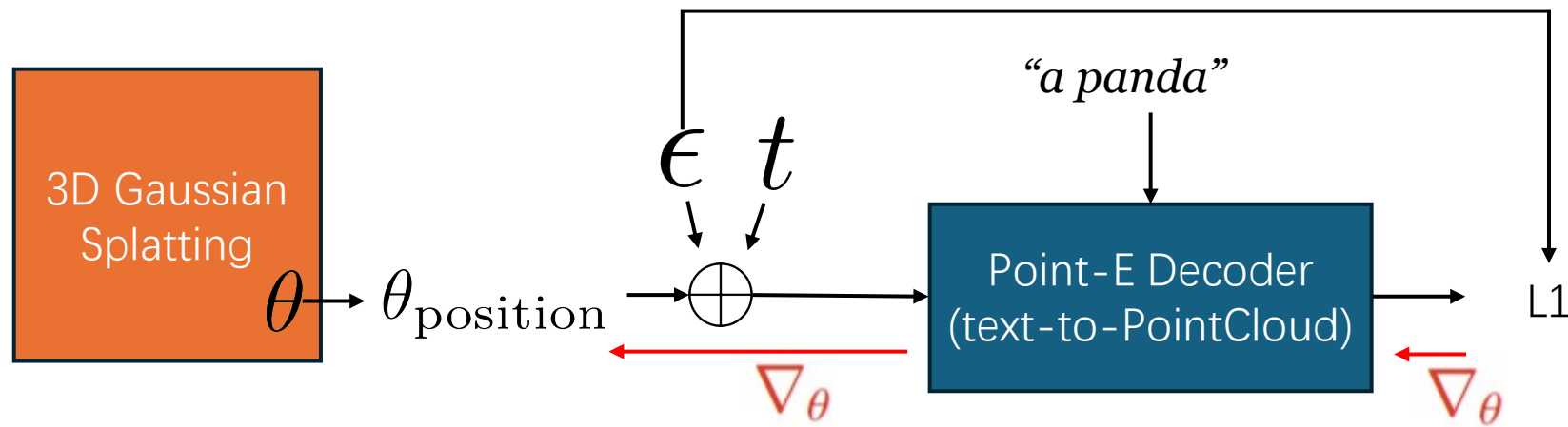
text to 3D

DreamFusion: Score Distillation Sampling, ICLR'23

Intuition: Novel-View Image = Noised Hidden Diffusion Step

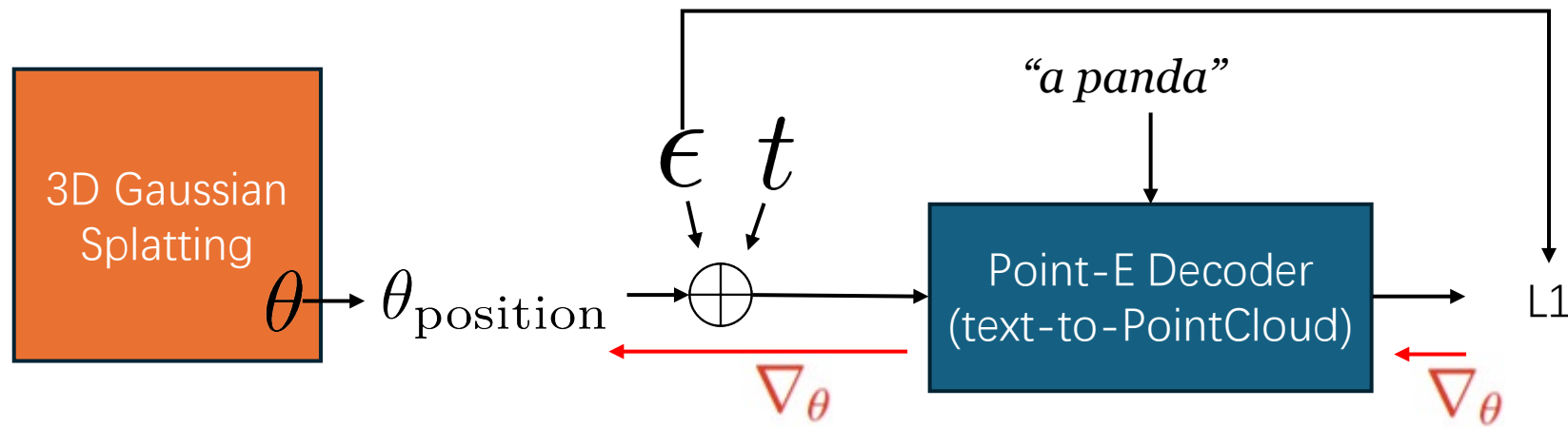


$$L_{\text{SDS}} = E_{t,\epsilon} \left[\left(DD(t, \epsilon, \mathbf{L}, \mathbf{x}_\phi) - \epsilon \right) \frac{\partial \mathbf{x}_\phi}{\partial \theta} \right]$$



$$L_{\text{PCD}} = E_{t, \epsilon} [(DD(t, \epsilon, \mathbf{L}, \theta_{\text{position}}) - \epsilon)]$$

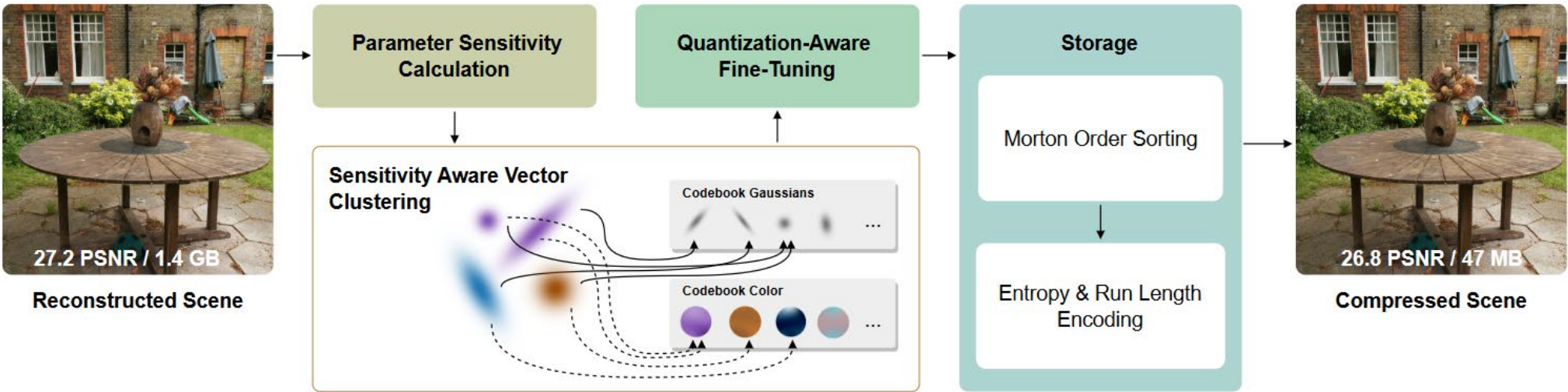
Intuition: Different from NeRF (implicit), GS are Explicit Point-based Representation



$$L_{\text{PCD}} = E_{t, \epsilon} [(DD(t, \epsilon, \mathbf{L}, \theta_{\text{position}}) - \epsilon)]$$

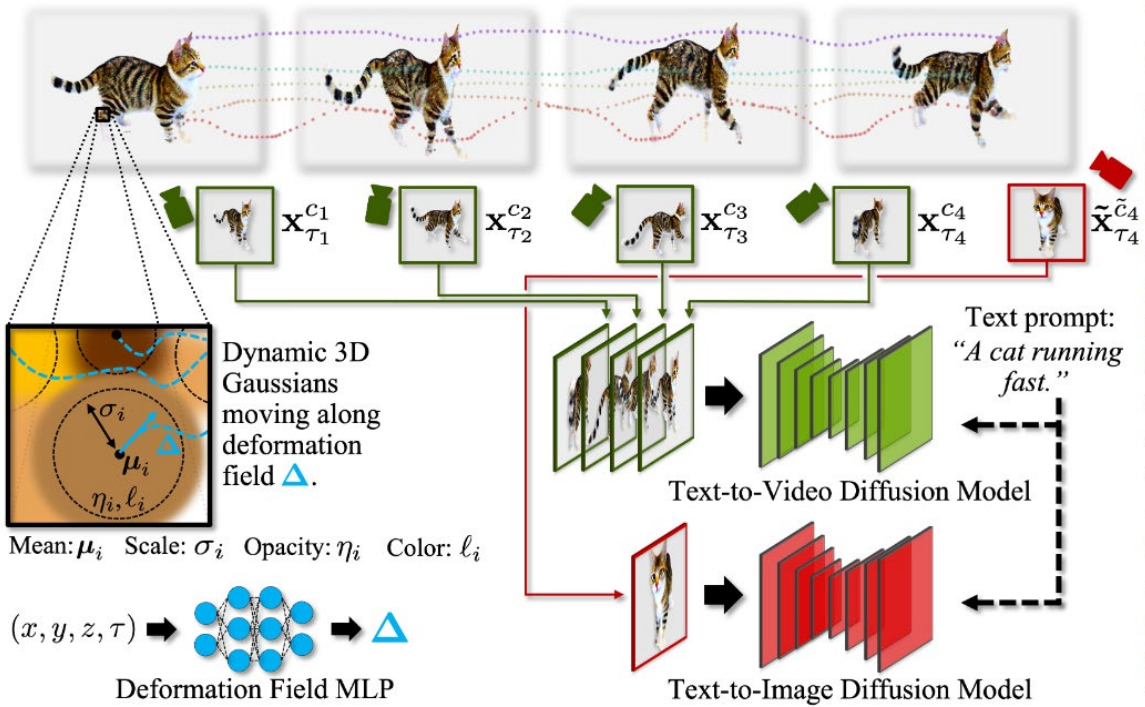
Efficient Gaussian Splatting

Compressed 3D Gaussian, CVPR'24



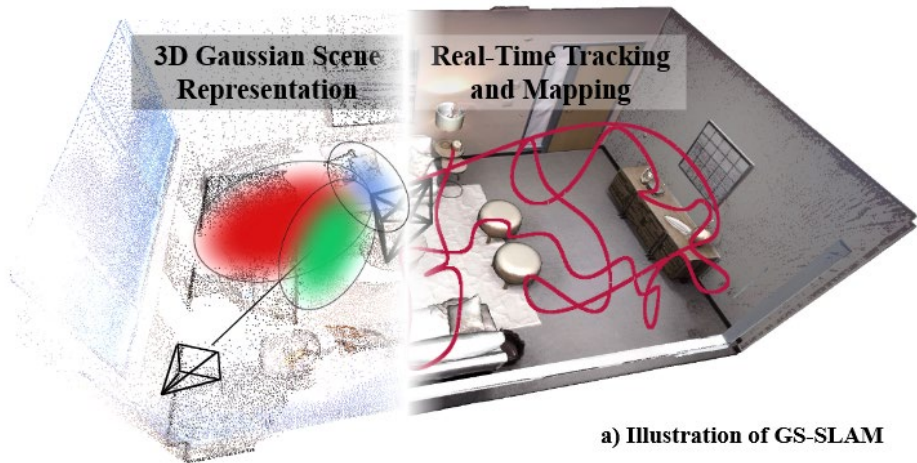
text to 4D

Align Your Gaussians, CVPR'24



Simultaneous Localization and Mapping

GS-SLAM, CVPR'24



Gaussian Splatting SLAM, CVPR'24



incremental 3D reconstruction using a single moving monocular or RGB-D camera