



Common Pets in 3D: Dynamic New-View Synthesis of Real-Life Deformable Categories

Samarth Sinha, Roman Shapovalov, Jeremy Reizenstein, Ignacio Rocco, Natalia Neverova, Andrea Vedaldi, David Novotny

Poster Session: Tuesday-PM-3772

https://cop3d.github.io/

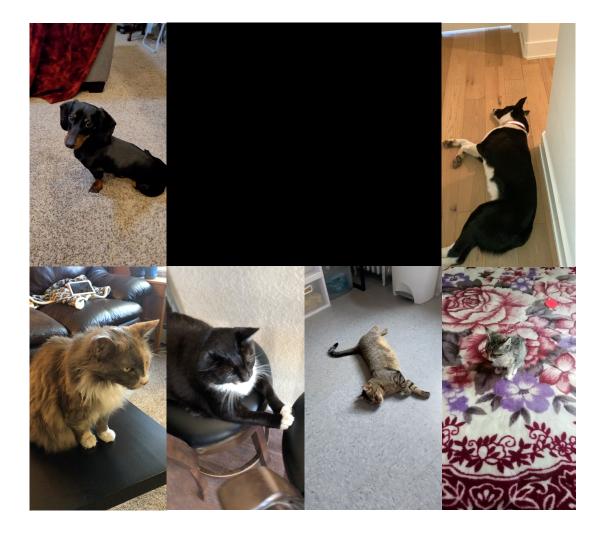


Common Pets in 3D Dataset: Overview



- First large-scale 3D dataset of dynamic objects
- Over 4,200 videos of cats and dogs
- Videos collected in-the-wild by crowdsourcing
- Collected with the camera panning around the pet
- Can enable category-centric 3D reconstruction

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Tracker NeRF $\Psi_{\rm CNN+CSE}(I_3^{\rm src}) \bigcirc$ $\Psi_{\text{CNN+CSE}}(I_2^{\text{src}})$ $I_3^{ m src}, t_3^{ m src}$ $I_2^{\rm src}, t_2^{\rm src}$ Rendering ray Target time t^{tgt} **ru**^{tgt} Target image I^{tgt} EA-raymarching Image render \bar{C}^{tgt} Point \mathbf{x} Source CNN+CSE Source time $t_1^{\rm src}$ dense features Photometric $\Psi_{\text{CNN+CSE}}(I_1^{\text{src}})$ Source image $I_1^{\rm src}$ loss $\mathcal{L}_{\mathrm{photo}}$ Adjusted point $\bar{\mathbf{x}}(t_1^{\mathrm{src}})$ CSE loss $\Psi_{\rm CSE}(I^{\rm tgt})$ $\mathcal{L}_{ ext{CSE}}$ Offset predicted Optical flow Projected adj. by NeRFormer Predicted CSE \bar{C}_{CSE}^{tgt} point $\bar{\mathbf{x}}(t_1^{\mathrm{src}})$ Optical flow loss $\delta(\mathbf{x}, t^{\text{tgt}}, t_1^{\text{src}})$ Rendered CSE $\mathcal{L}_{\mathrm{flow}}$

- Learns category-specific priors from a large-scale dataset
- Pretrained prior provides good initialization for single-scene overfitting

Experiments: Reconstructions







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3D Reconstruction Datasets



Dynamic NeRF datasets

- Easy to collect and annotate
- Static and Dynamic objects
- Only single scenes
- Used only for overfitting

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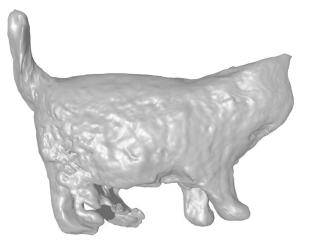
Common Objects in 3D (Reizenstein et al.)

- Large scale 3D dataset
- Enables learning category priors
- Only static objects
- Challenging to collect and annotate

Non-Rigid 3D Reconstruction





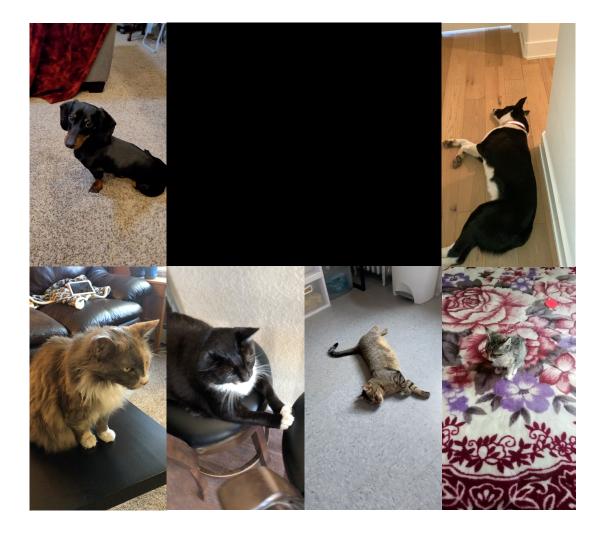


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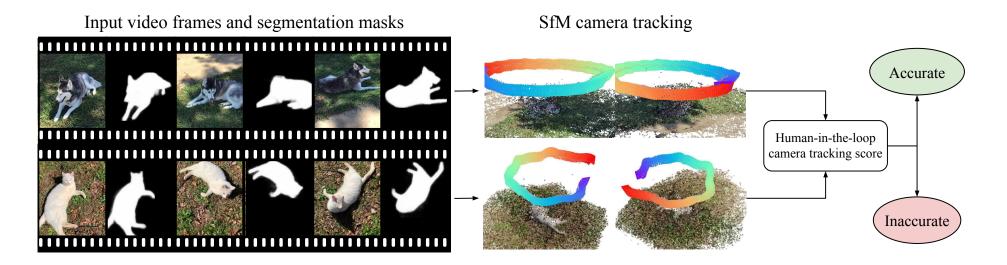
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Common Pets in 3D Dataset: Annotation



Full 3D annotation available including:

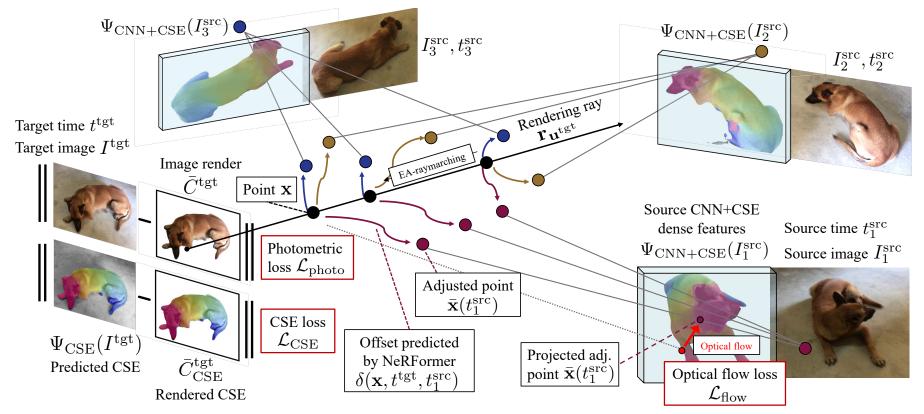
- 200 frames per scene
- 3D camera trajectory
- Camera quality scores
- Segmentation masks

Tracker NeRF: Overview



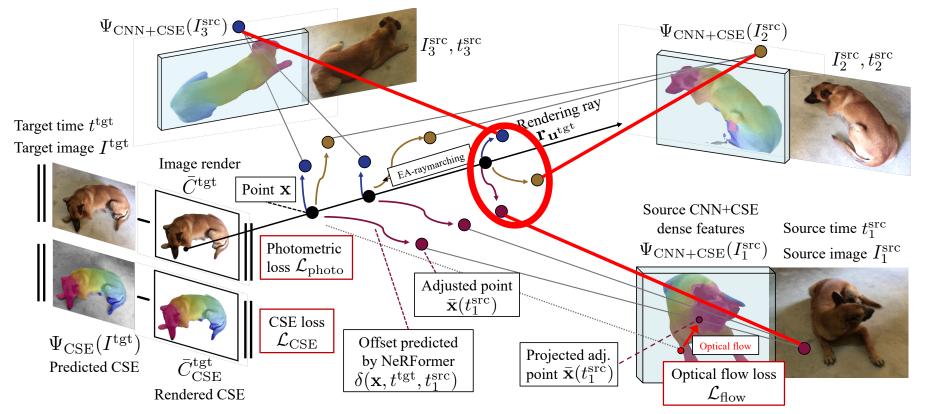
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Tracker NeRF: Method



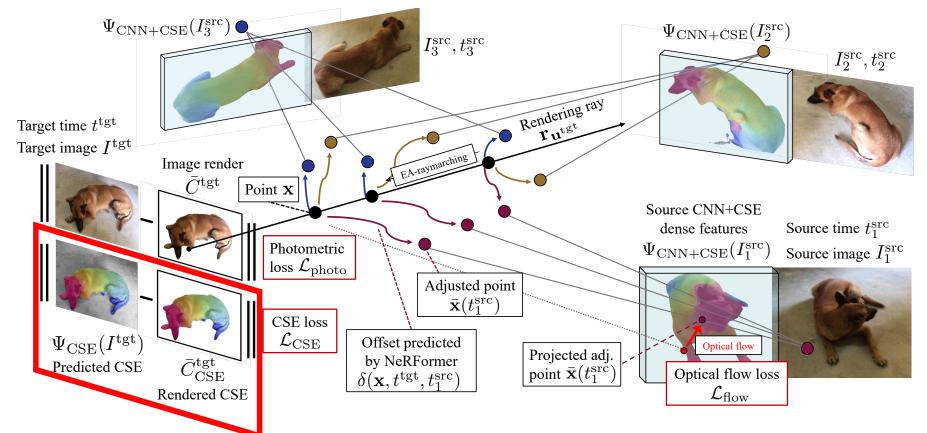
- Tracker NeRF utilizes a NeRFormer as its base architecture (Reizenstein et al., ICCV 2021)
- Tracker NeRF utilizes supervision from 2D optical flow to learn dynamic priors and Continuous Surface Embeddings (CSEs) to learn priors over the category

Tracker NeRF: Method



- For a point sampled along the target ray, Tracker NeRF predicts the 3D offsets for each source view
- That point is then projected into the source views and the features are sampled at the new location
- The prediction is supervised using 2D optical flow

Tracker NeRF: Method

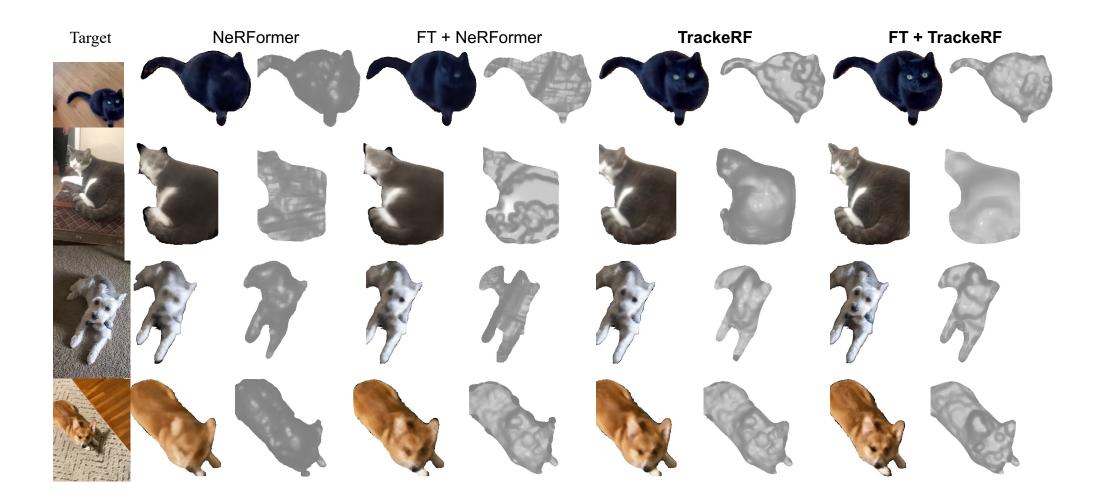


- Tracker NeRF also predicts the surface embedding for each 3D point
- The surface embedding can then be rendered and supervised by the ground-truth embedding for the target views

Experiments: Single-scene Overfitting

| \downarrow Method | PSNR | LPIPS | IoU | $\ell_1^{\rm RGB}$ |
|---------------------|-------------|-------------|------|--------------------|
| SRN+AD | 17.2 | 0.24 | 0.78 | 0.27 |
| SRN+TWCE | 16.8 | 0.19 | 0.75 | 0.29 |
| TimeNeRF | 17.3 | 0.18 | 0.72 | 0.20 |
| NeRF+TWCE | 17.3 | 0.19 | 0.73 | 0.46 |
| NeRFormer+TWCE | 18.6 | 0.17 | 0.82 | 0.21 |
| NSFF | 20.2 | 0.17 | | <u>0.19</u> |
| TrackeRF (ours) | <u>21.4</u> | <u>0.15</u> | 0.91 | 0.17 |
| FT+NeRF+TWCE | 17.7 | 0.19 | 0.82 | 0.30 |
| FT+NeRFormer+TWCE | 20.5 | <u>0.15</u> | 0.88 | 0.20 |
| FT+TrackeRF (ours) | 23.1 | 0.13 | 0.91 | 0.17 |

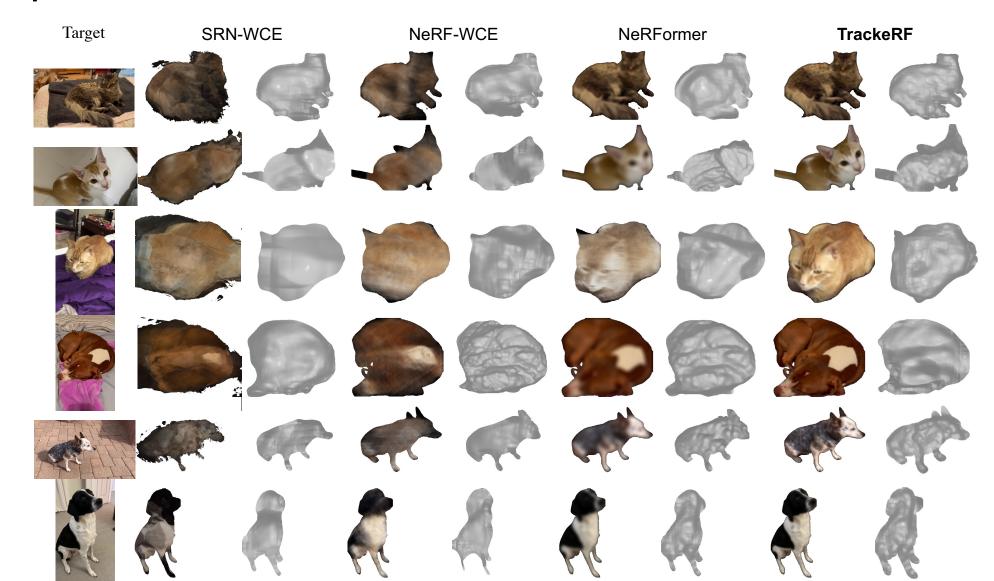
Experiments: Single-scene Overfitting



Experiments: Category Reconstruction

| | (a) Average statistics | | | | | (b) PSNR @ # source views | | | | | | | | | | | | |
|-------------------------|------------------------|-------|----------------------|-------------|------|---------------------------|----------------------|------|------|------|------|-------------|------|------|------|------|------|------|
| Frame set \rightarrow | train-unseen | | | test-unseen | | | train-unseen | | | | | test-unseen | | | | | | |
| \downarrow Method | PSNR | LPIPS | $\ell_1^{	ext{RGB}}$ | IoU | PSNR | LPIPS | $\ell_1^{	ext{RGB}}$ | IoU | 25 | 20 | 15 | 10 | 5 | 25 | 20 | 15 | 10 | 5 |
| TrackeRF (ours) | 19.1 | 0.16 | 0.33 | 0.80 | 19.6 | 0.17 | 0.31 | 0.82 | 19.7 | 19.6 | 19.6 | 18.8 | 17.6 | 21.0 | 20.0 | 20.0 | 18.2 | 17.9 |
| NeRFormer+TWCE [38] | 16.3 | 0.19 | 0.39 | 0.73 | 16.6 | 0.18 | 0.36 | 0.76 | 16.5 | 17.2 | 16.6 | 16.3 | 14.9 | 17.7 | 16.9 | 17.0 | 15.6 | 15.7 |
| NeRF+TWCE [10] | 14.6 | 0.20 | 0.48 | 0.60 | 14.2 | 0.20 | 0.44 | 0.60 | 14.8 | 15.1 | 14.6 | 14.6 | 14.0 | 15.9 | 15.3 | 15.0 | 14.4 | 15.0 |
| SRN+TWCE | 13.9 | 0.18 | 0.52 | 0.53 | 14.2 | 0.18 | 0.49 | 0.53 | 13.7 | 14.6 | 14.3 | 13.7 | 13.1 | 15.0 | 14.4 | 14.3 | 13.3 | 14.2 |
| SRN+AD [43] | 15.5 | 0.19 | 0.40 | 0.66 | | - | | | 15.0 | 16.2 | 15.5 | 15.1 | | | - | | | |

Experiments: Category Reconstruction



Experiments: Reconstructions



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