



NeRF with Optical Satellite Images

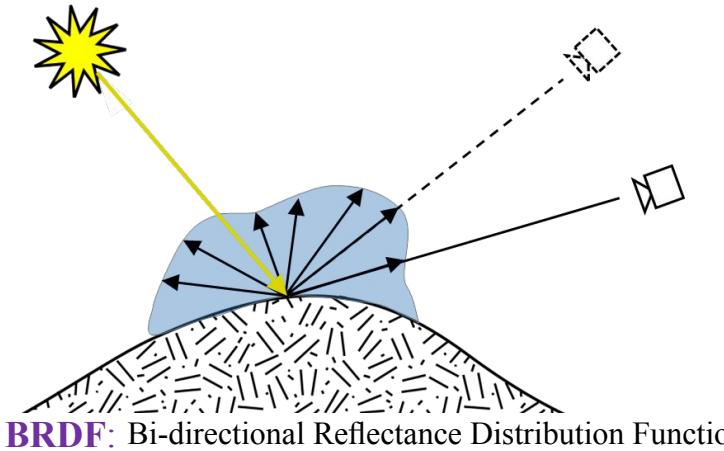
(NeRF: Neural Radiance Fields)

Lulin Zhang (IPGP, LASTIG)

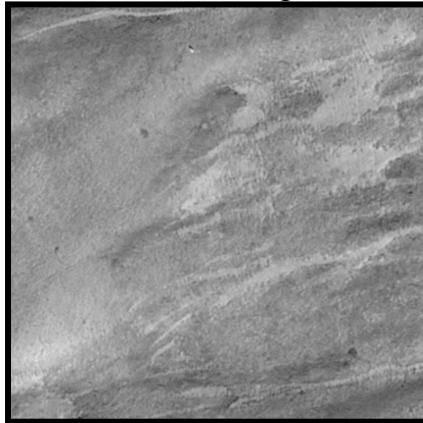
Supervisors: Ewelina Rupnik (LASTIG)
Yann Klinger (IPGP)
Marc Pierrot-Deseilligny (LASTIG)

In collaboration with: Stéphane Jacquemoud (IPGP)

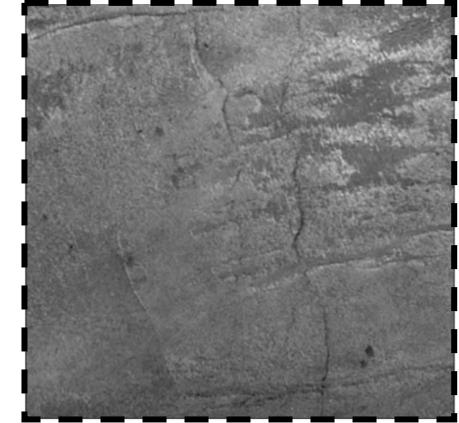
Context



before earthquake

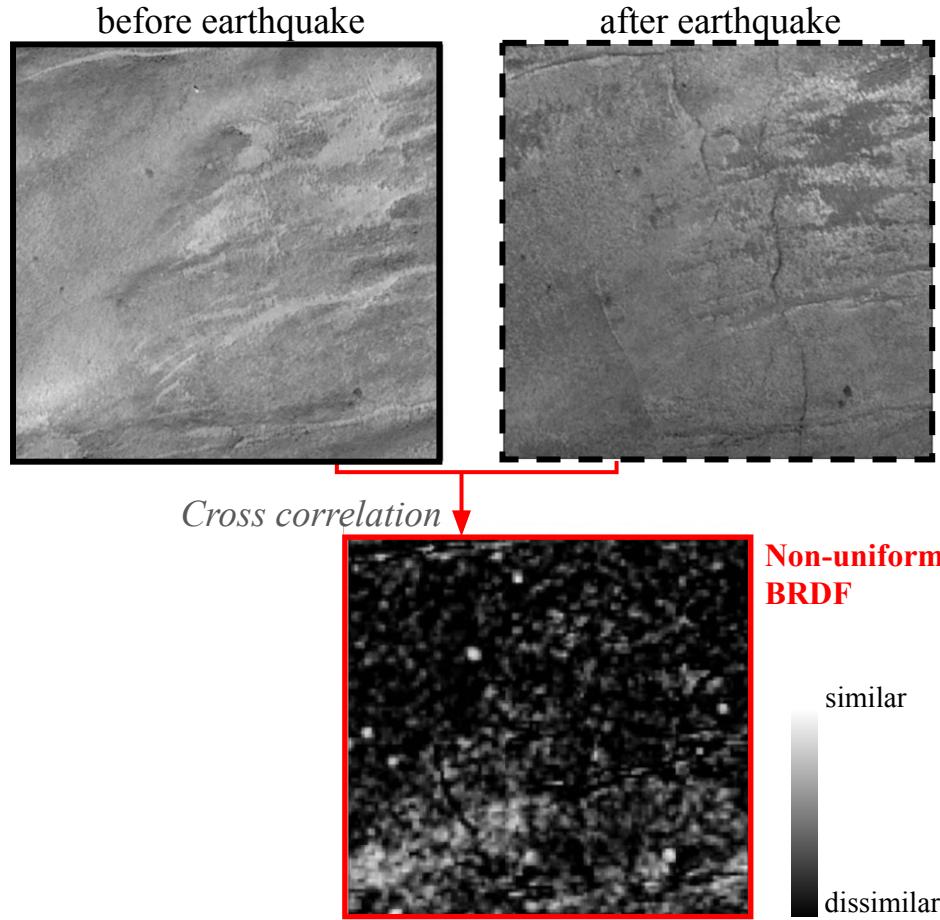
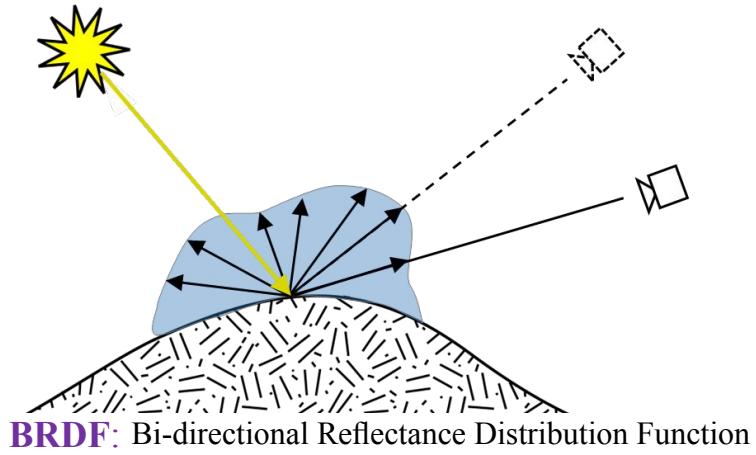


after earthquake



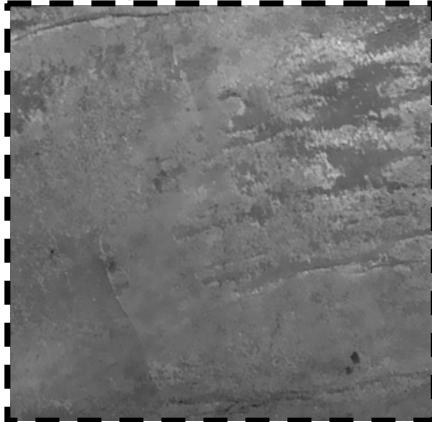
BRDF: Bi-directional Reflectance Distribution Function

Context

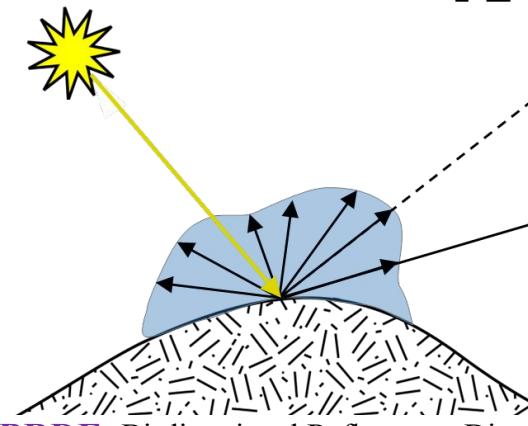
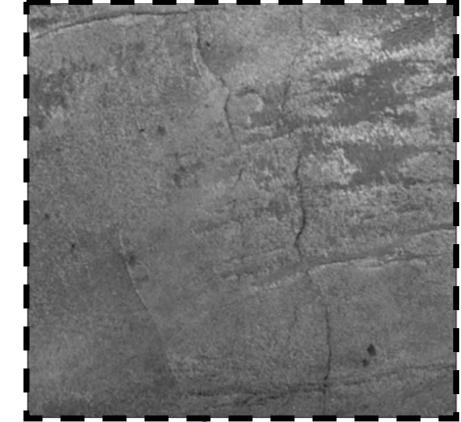


Context

Synthetic image

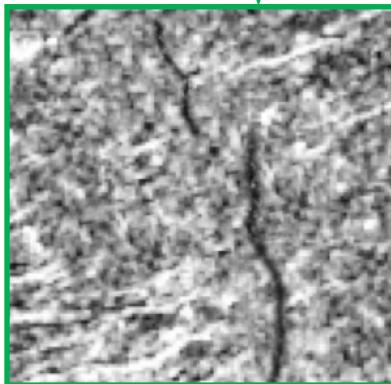


after earthquake



Uniform
BRDF

Cross correlation



similar

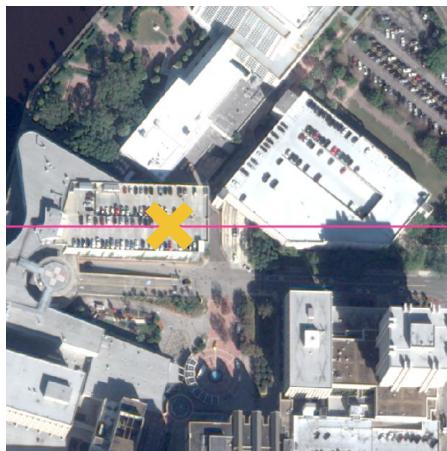


BRDF: Bi-directional Reflectance Distribution Function

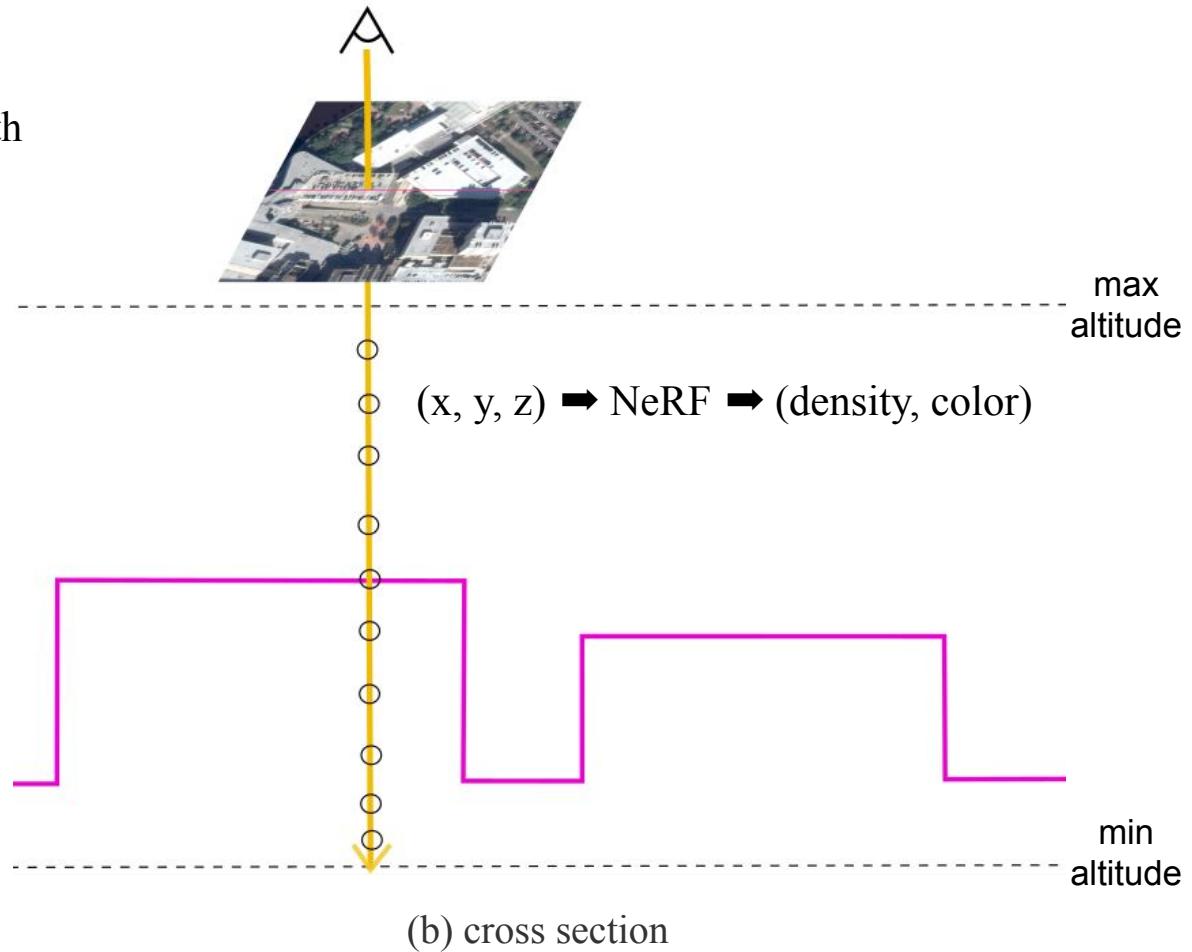
NeRF (Neural Radiance Fields)

NeRF:

- Deep learning.
- Treat the scene as a volume with density and color.
- Lots of variants.



(a) A selected image row (-)

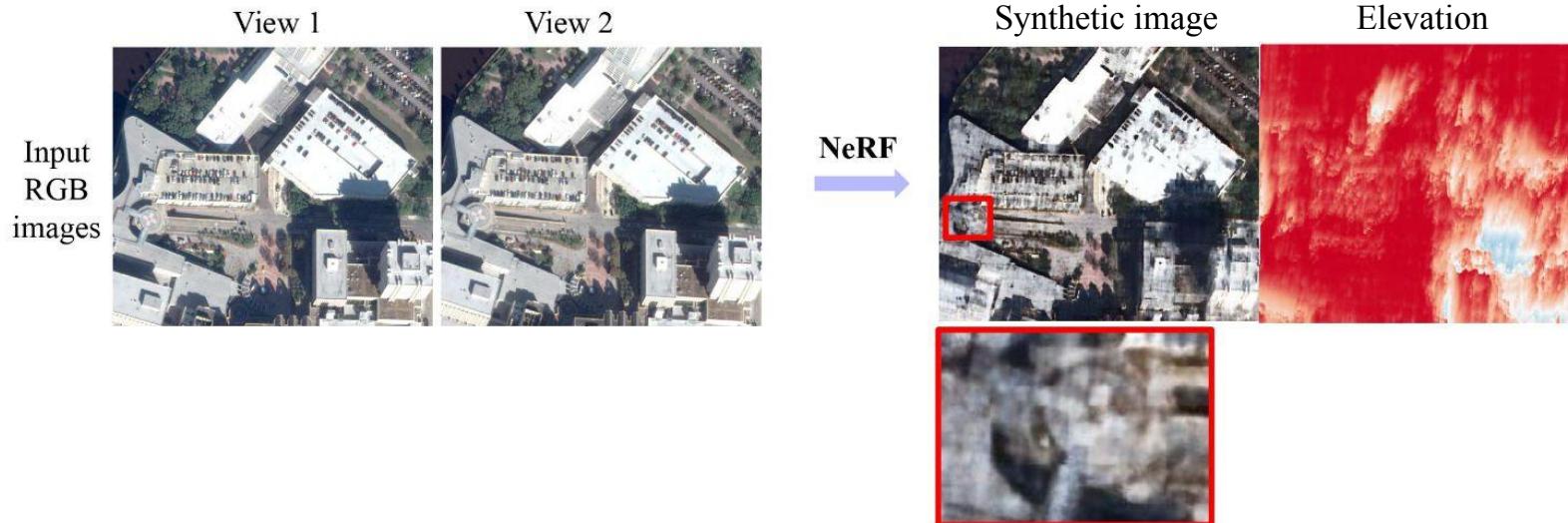


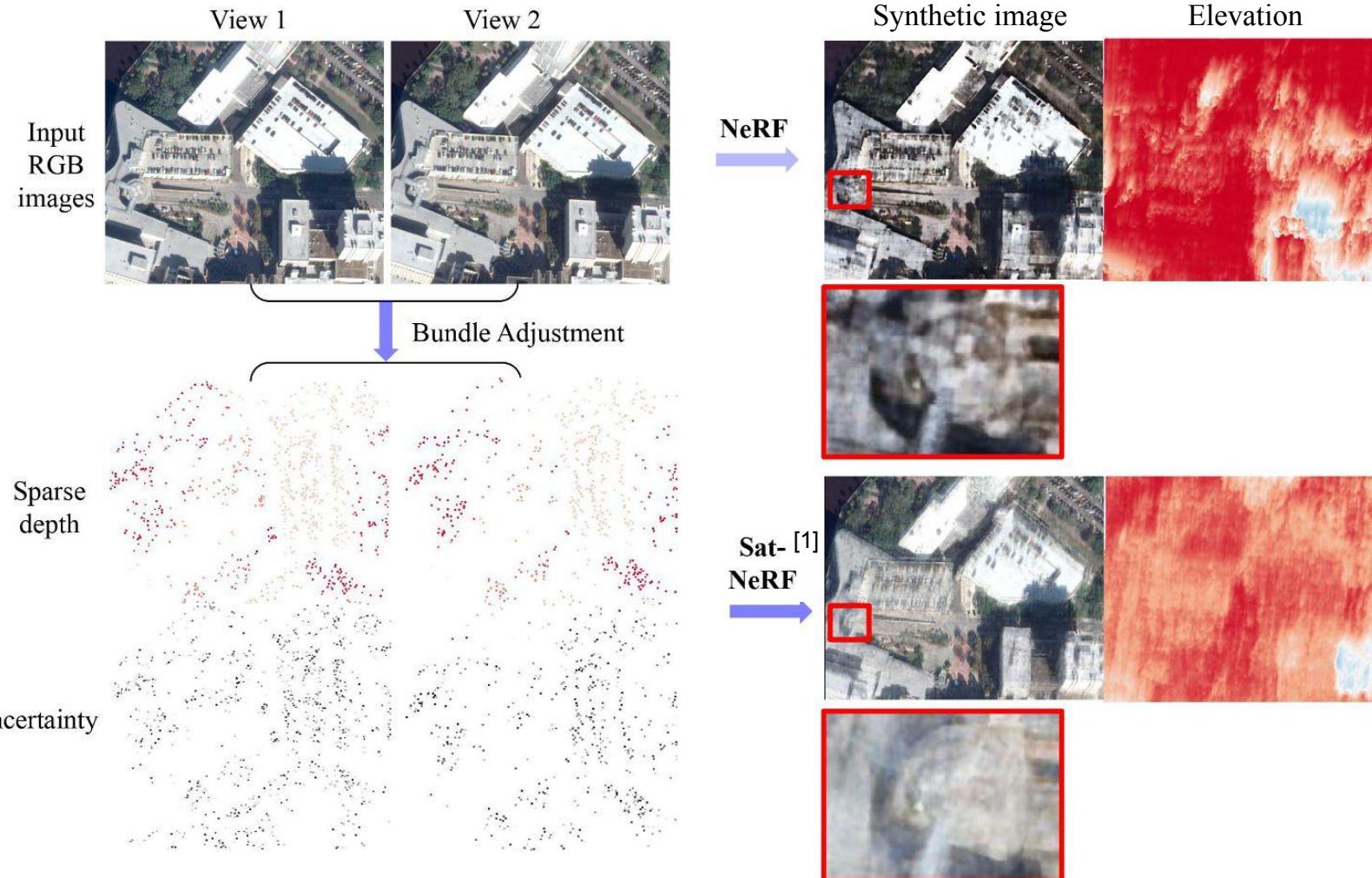
Challenges

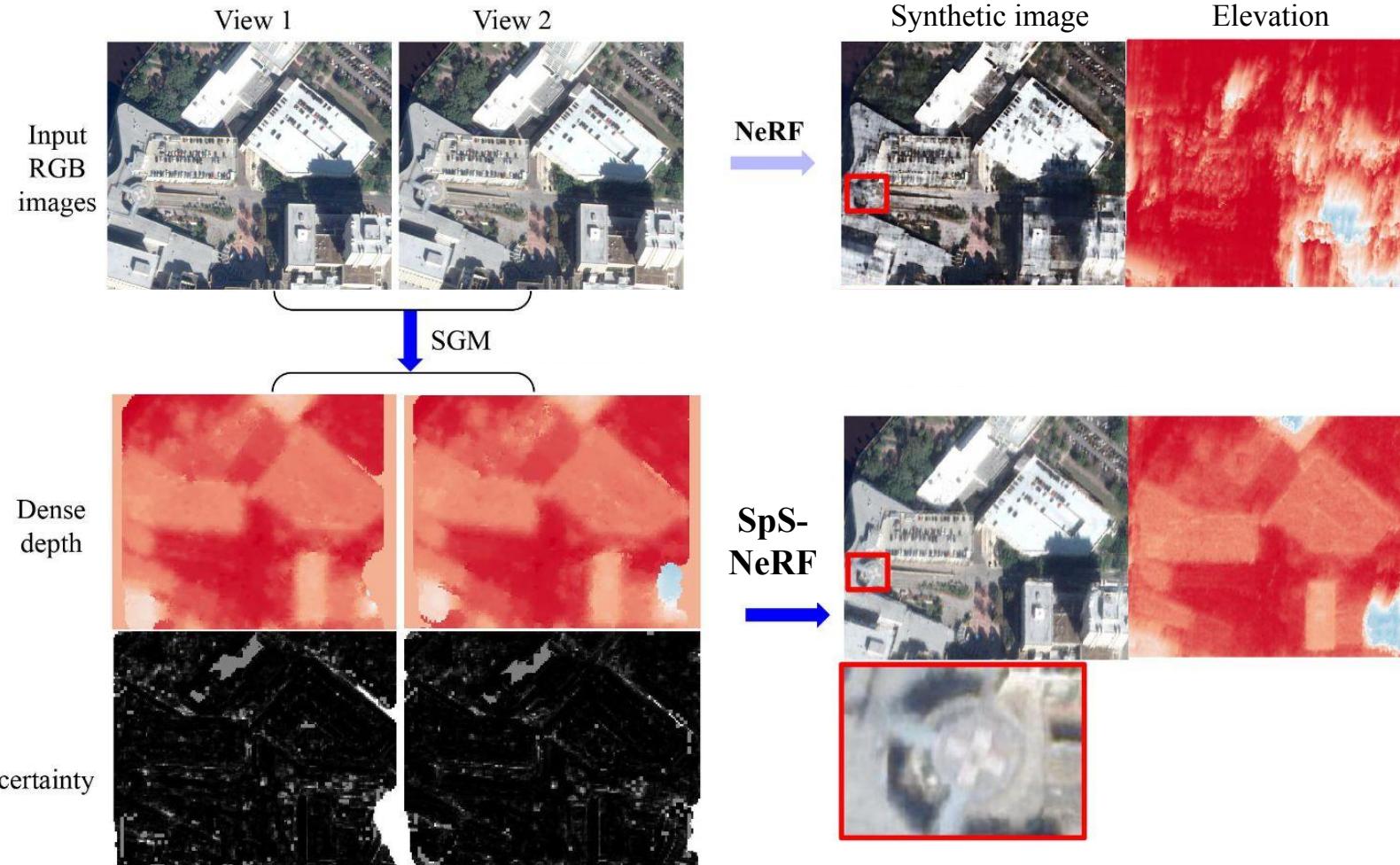
Two big challenges in our work:

1. Sparse input images

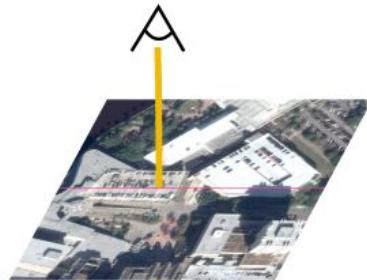




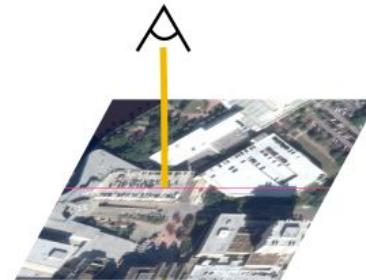




SpS-NeRF: Guided sampling



Stratified sampling



Guided sampling

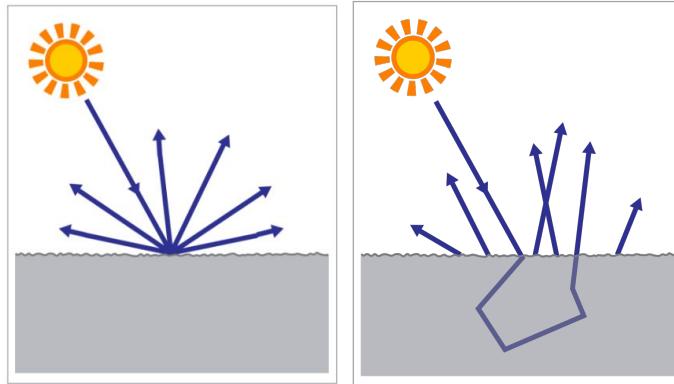
Challenges

Two big challenges in our work:

1. Sparse input images

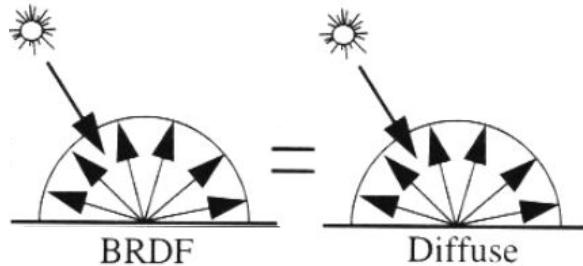


2. Complex BRDF in remote sensing images



BRDF model

Lambertian:



Reflectance:

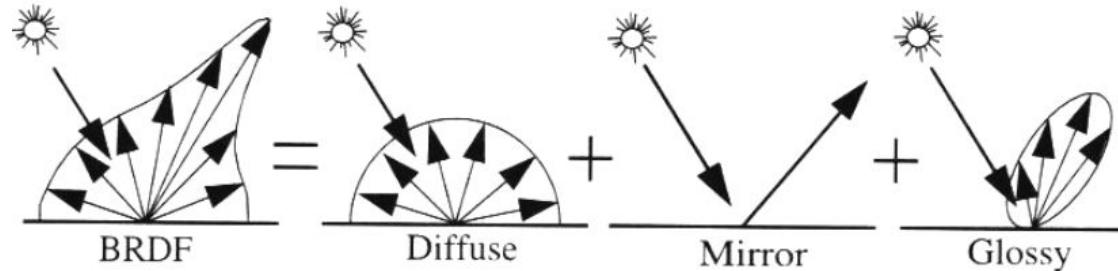
- All the light is reflected **equally in all directions**.

Example:



BRDF model

Microfacet:



Reflectance:

- Some light is reflected **equally in all directions**.
- The rest is reflected **near the specular direction**.

Example:



BRDF model

RPV: $BRF(\theta_i, \theta_e, \phi, \lambda) = \rho_0(\lambda) \times M_I(\theta_i, \theta_e, k) \times F_{HG}(g, \Theta) \times H(\rho_c, G)$

Hapke: $r(i, e, g) = \frac{w}{4\pi} \frac{\mu_{0e}}{\mu_{0e} + \mu_e} \left\{ p(g, b, c) + H(\mu_{0e})H(\mu_e) - 1 \right\} S(i, e, g, \theta)$

Reflectance:

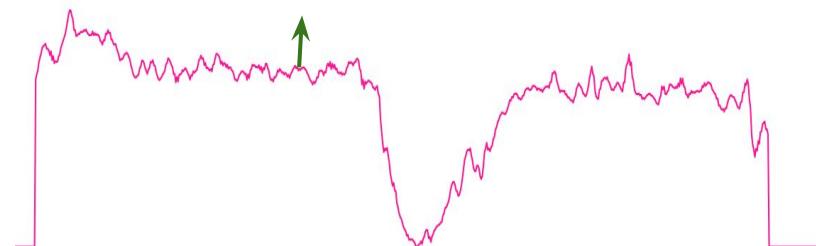
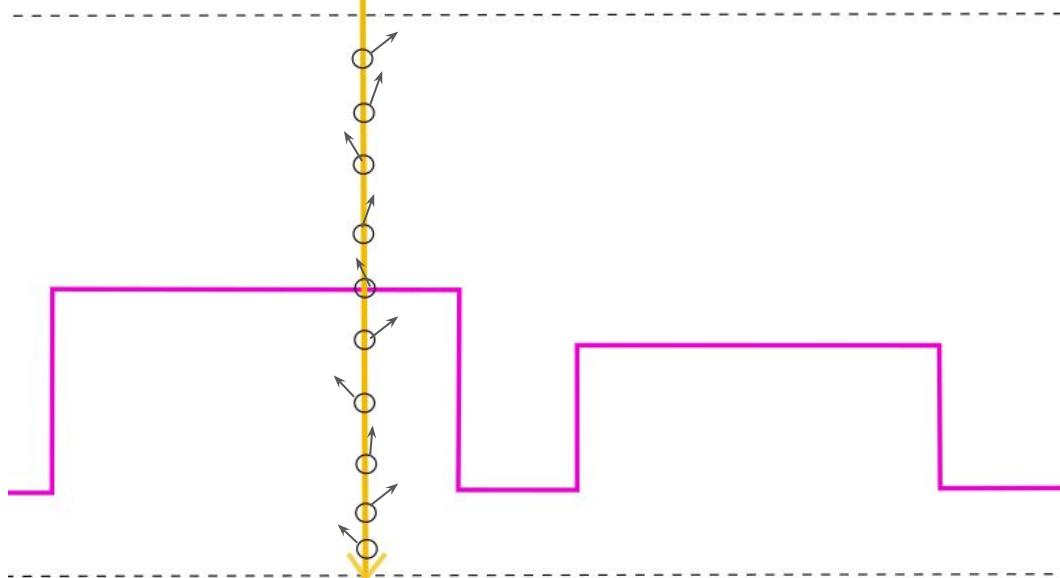
- Some light is reflected **equally in all direction**.
- The rest energy can be reflected **in arbitrary direction with arbitrary amplitude**, depending on parameters.
- Suitable for arbitrary natural surfaces, including bare soils and vegetation.

Example:



Normal supervision (nr)

Surface normal is essential in estimating BRDF.

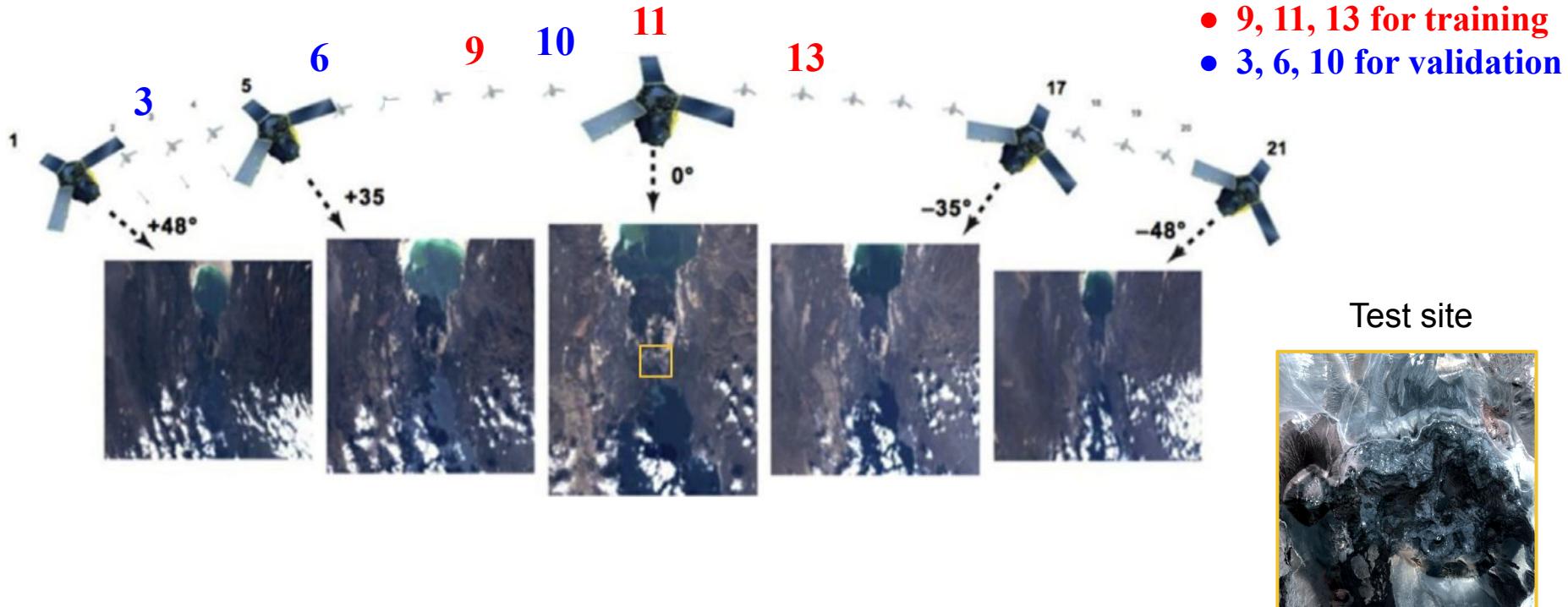


Normal from dense depth

Normal from network

Experiments

Djibouti dataset. It contains 21 multiangular Pléiades images collected in a single flyby.



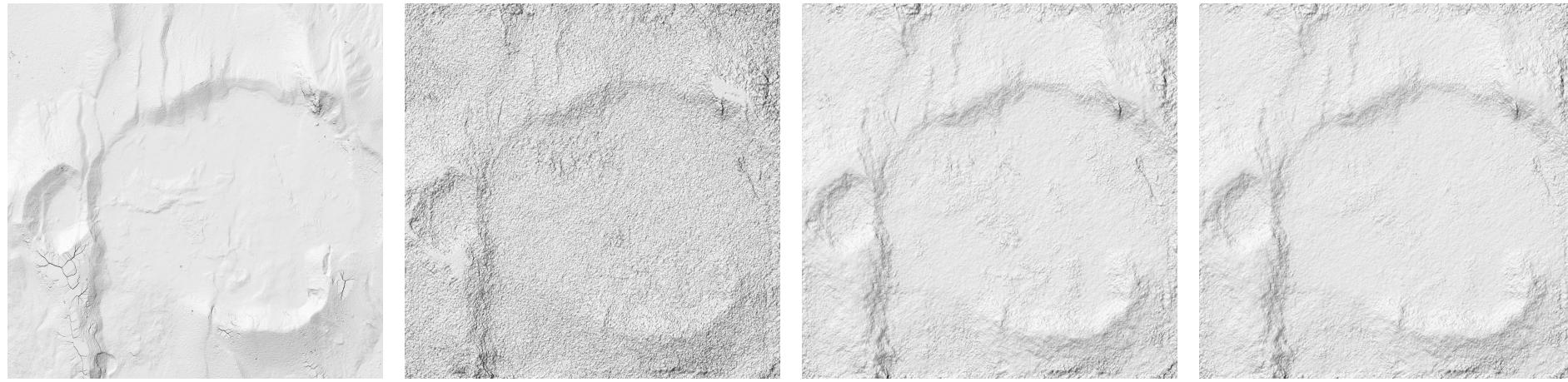
(1) Elevation:



NeRF

Sat-NeRF

SGM

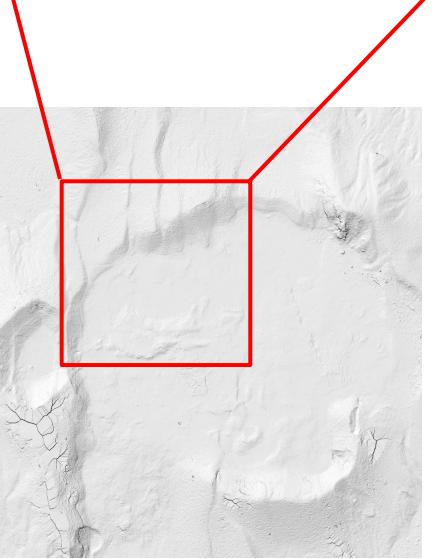
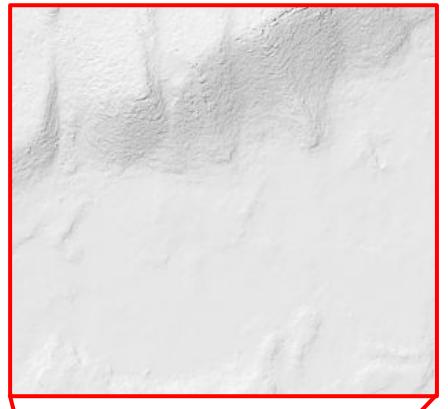


GT

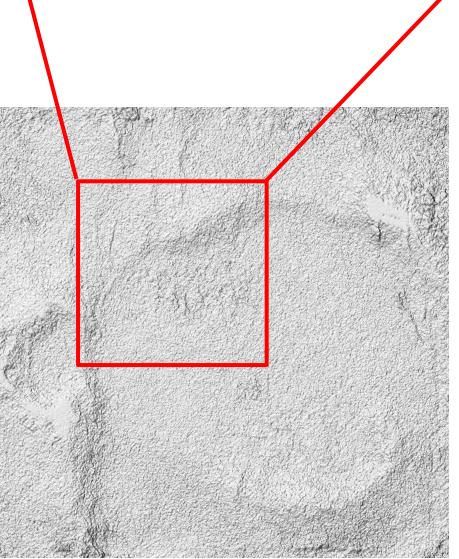
SpS-NeRF

SpS-NeRF + nr

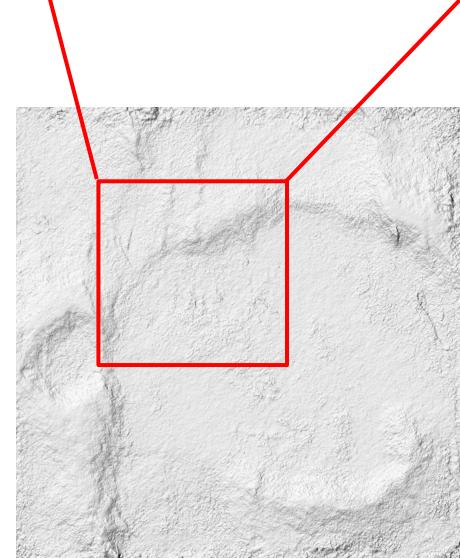
SpS-NeRF + nr + RPV



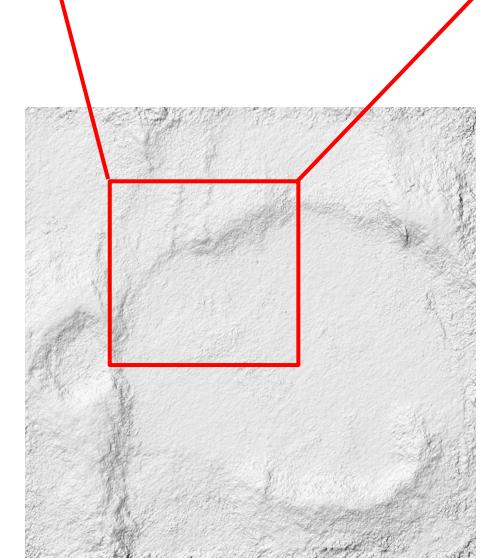
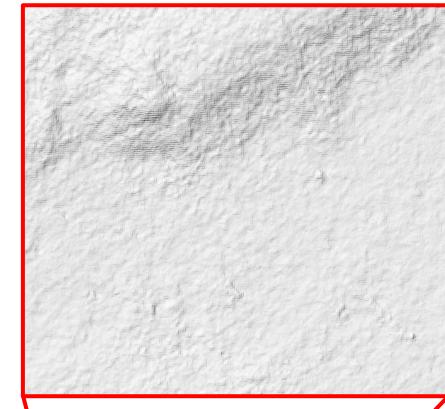
GT



SpS-NeRF

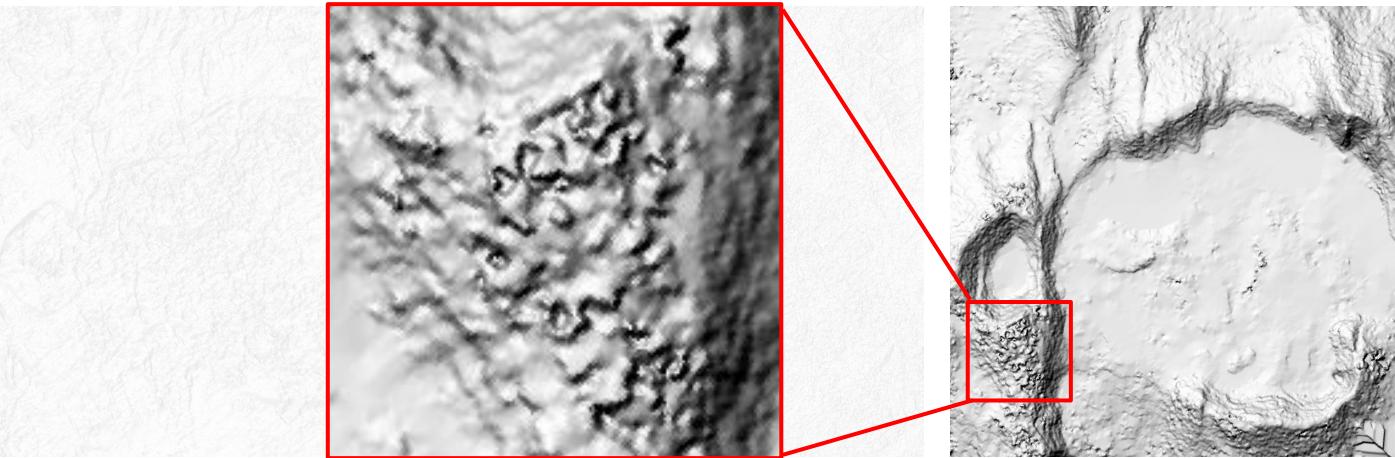


SpS-NeRF + nr



SpS-NeRF + nr + RPV

(1) Elevation:



NeRF

Sat-NeRF

SGM

GT

SpS-NeRF

SpS-NeRF + nr

SpS-NeRF + nr + RPV

(2) Synthetic image:



NeRF



Sat-NeRF



SGM



GT



SpS-NeRF



SpS-NeRF + nr



SpS-NeRF + nr + RPV

(2) Synthetic image:



NeRF



Sat-NeRF



SGM



GT



SpS-NeRF



SpS-NeRF + nr



SpS-NeRF + nr + RPV

(2) Synthetic image:



Quantitative metrics:

	PSNR 	SSIM 	MAE 
NeRF	29.24	0.83	13.38
Sat-NeRF	29.65	0.85	10.11
SpS-NeRF	31.47	0.91	2.09
SpS-NeRF + nr	31.72	0.91	1.52
SpS-NeRF + nr + RPV	31.81	0.93	1.36
SGM	/	/	0.81

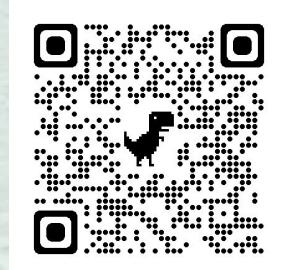
*mean value on 3 validation images.



Thank you
for your attention !



Code



Project page



Email me