



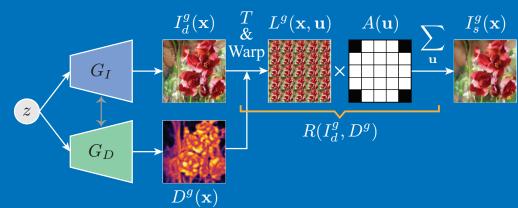


Project page

Unsupervised Learning of Depth and Depth-of-Field Effect from Natural Images with Aperture Rendering Generative Adversarial Networks



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Overview



Unsupervised learning of depth and DoF effect from natural images

Unlabeled natural images



AR-GAN



Deep DoF Shallow DoF Depth



Training data

- No ground-truth depth
- No paired data
- No pretrained model

Generated data

$$G: z \mapsto (I_d^g, I_s^g, D^g)$$

Positioning of research



Fully unsupervised 3D representation learning

Only a collection of unlabeled natural images are available for training

Viewpoint-aware

Learn 3D representation using *viewpoint* cues

HoloGAN

[Nguyen-Phuoc+2019]

Szabó et al.

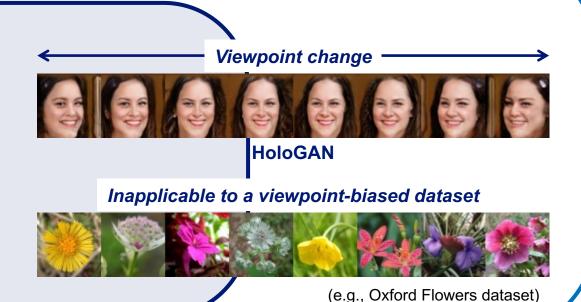
[Szabó+2019]

RGBD-GAN

[Noguchi+2020]

Unsup3d

[Wu+2020]



Positioning of research



Fully unsupervised 3D representation learning

Only a collection of unlabeled natural images are available for training

Focus-aware

Learn 3D representation using focus cues

Applicable to a viewpoint-biased dataset if the dataset includes various DoF images







Training data (e.g., Oxford Flowers dataset)













AR-GAN

[ours]

Positioning of research



Fully unsupervised 3D representation learning

Only a collection of unlabeled natural images are available for training

Viewpoint-aware

Learn 3D representation using *viewpoint* cues

HoloGAN

[Nguyen-Phuoc+2019]

Szabó et al. [Szabó+2019]

RGBD-GAN

[Noguchi+2020]

Unsup3d [Wu+2020]

AR-**HoloGAN** [ours] Compatible AR-**RGBD-GAN** [ours]

Learn 3D representation using *focus* cues

Focus-aware

AR-GAN

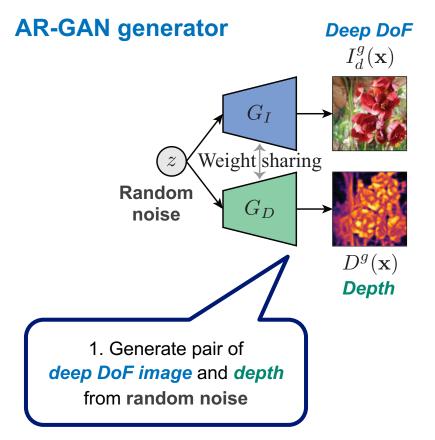
[ours]

Viewpoint change



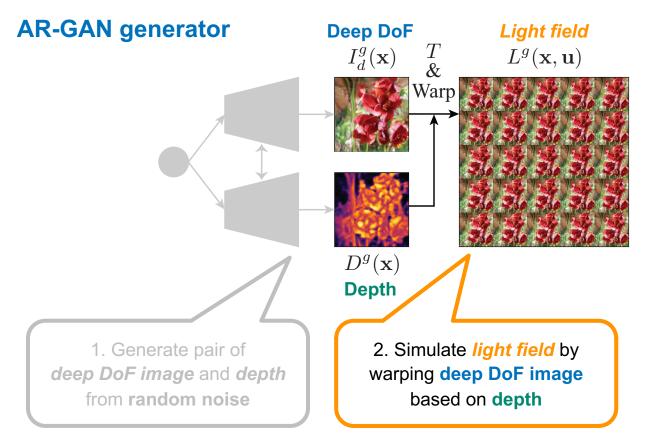
Overall pipeline 1/3





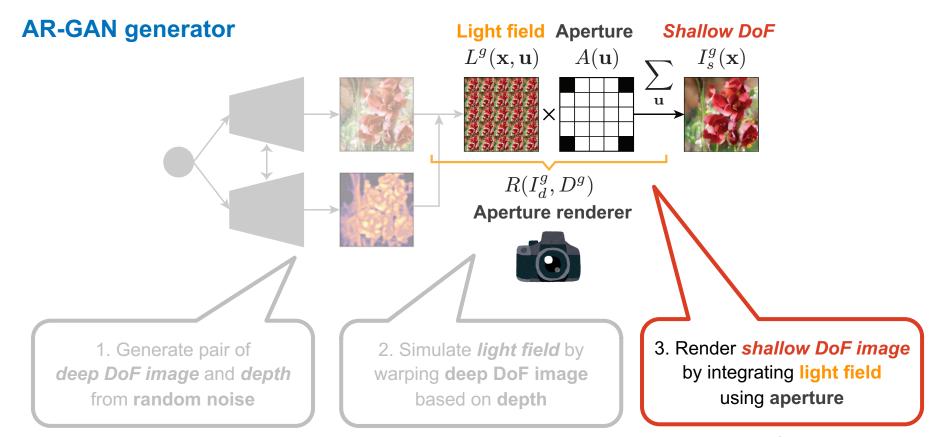
Overall pipeline 2/3





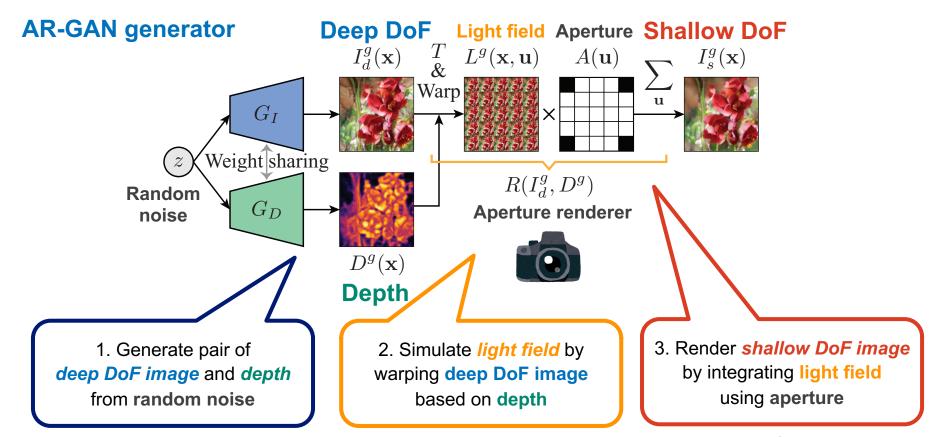
Overall pipeline 3/3





Overall pipeline

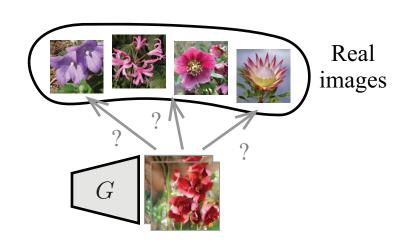




Standard learning (baseline)



Learn real image distribution without any constraints



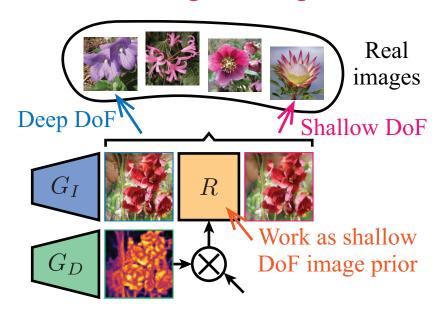
Standard learning (baseline)

$$\mathcal{L}_{GAN} = \mathbb{E}_{I^r}[\log C(I^r)] + \mathbb{E}_z[\log(1 - C(G(z)))]$$

DoF mixture learning (proposed)



Learn real image distribution while generating diverse DoF images



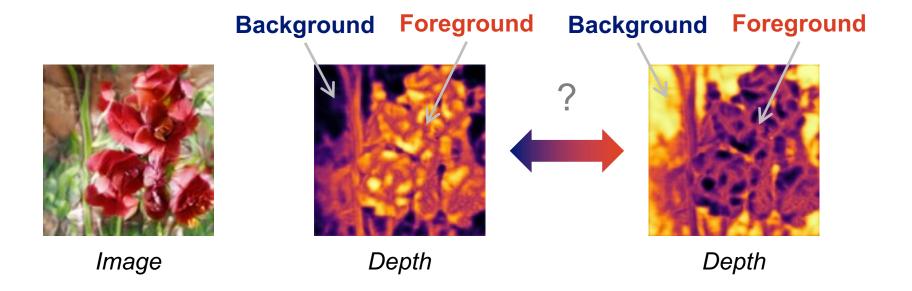
DoF mixture learning (proposed)

$$\mathcal{L}_{AR-GAN} = \mathbb{E}_{I^r}[\log C(I^r)] + \mathbb{E}_{z,s}[\log(1 - C(R(G_I(z), sG_D(z)))]$$

Difficulty in unsupervised learning



Ambiguity between fore/background blur



Center focus prior



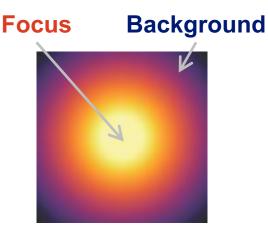
Solve ambiguity between fore/background blur by providing prior







Examples of center focused images

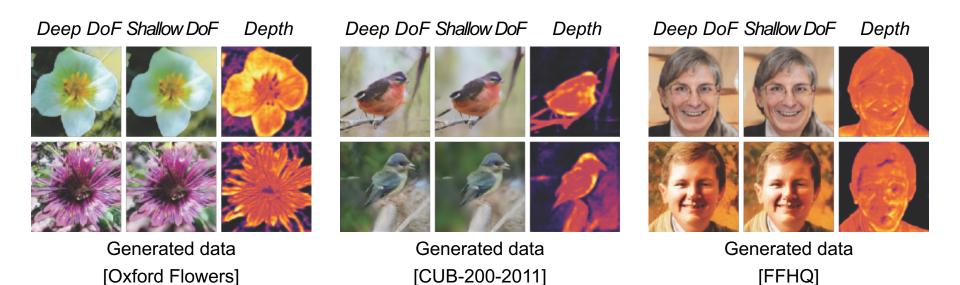


Center focus prior

Examples of generated data



Examples of data generated from AR-GAN

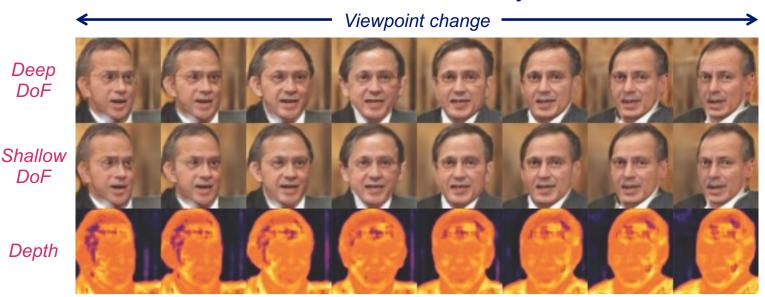


Portability analysis



Examples of data generated from AR-RGBD-GAN (AR-GAN + RGBD-GAN)

RGBD-GAN functionality



Generated data [FFHQ]

AR-GAN functionality

DoF

DoF

Application in shallow DoF rendering



Learn shallow DoF renderer (Deep DoF → **Shallow DoF) using generated data**





Input (iPhone photo)

Shallow DoF



Depth



AR-GAN-DR

(proposed)

W/ **no** supervision

Shallow DoF



CycleGAN (baseline)

W/ set-level supervision

Deep or shallow DoF





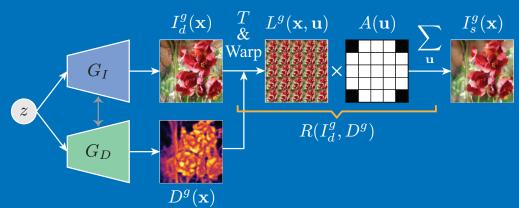


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