

# Deep Compressive Sensing for Visual Privacy Protection in FlatCam Imaging

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### 1. Introduction

Detection then protection is vulnerable to

**LCI** 

Workshop

- FlatCam<sup>[1,2]</sup> captures visually protected but not privacy protected image & low recovery quality.
- We improved reconstruction with a deep residual learning network.
- We protect sensitive regions from FlatCam

# **DECODER ENCODER Authorized User** $X \rightarrow Sample \rightarrow Recovery \rightarrow Face Seg. \rightarrow \Longrightarrow \rightarrow$ Normal User Kronecker sampling $Y \approx PXQ^T$

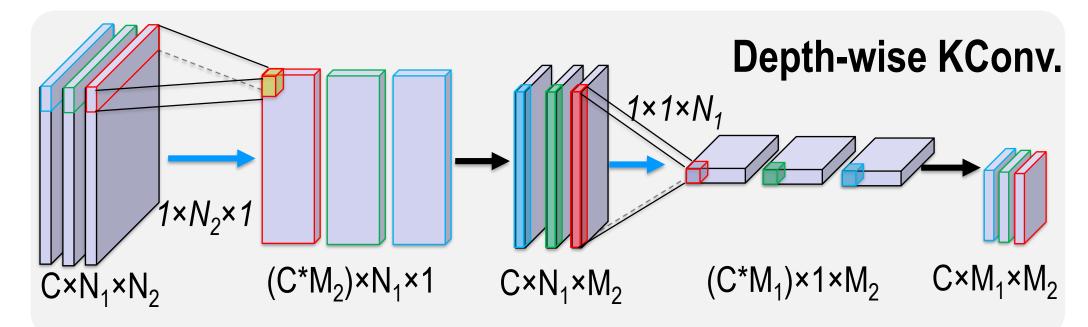
Proposed Deep Visual Privacy Protection Framework

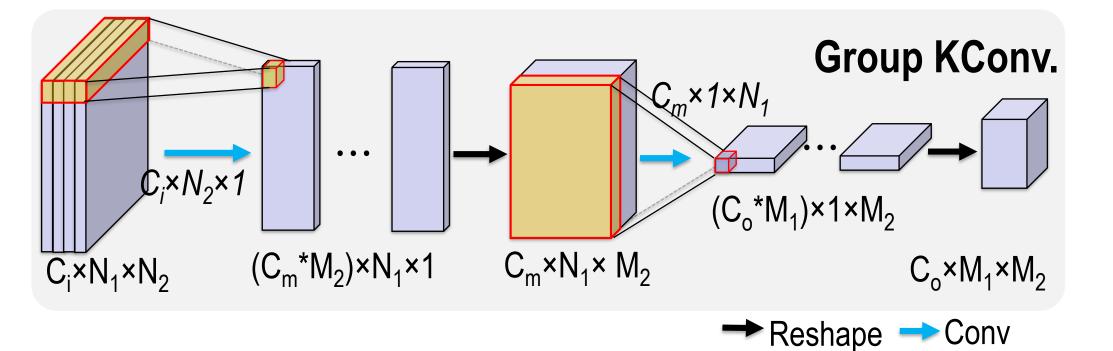
- software attack.

- measurement via a deep visual privacy network.

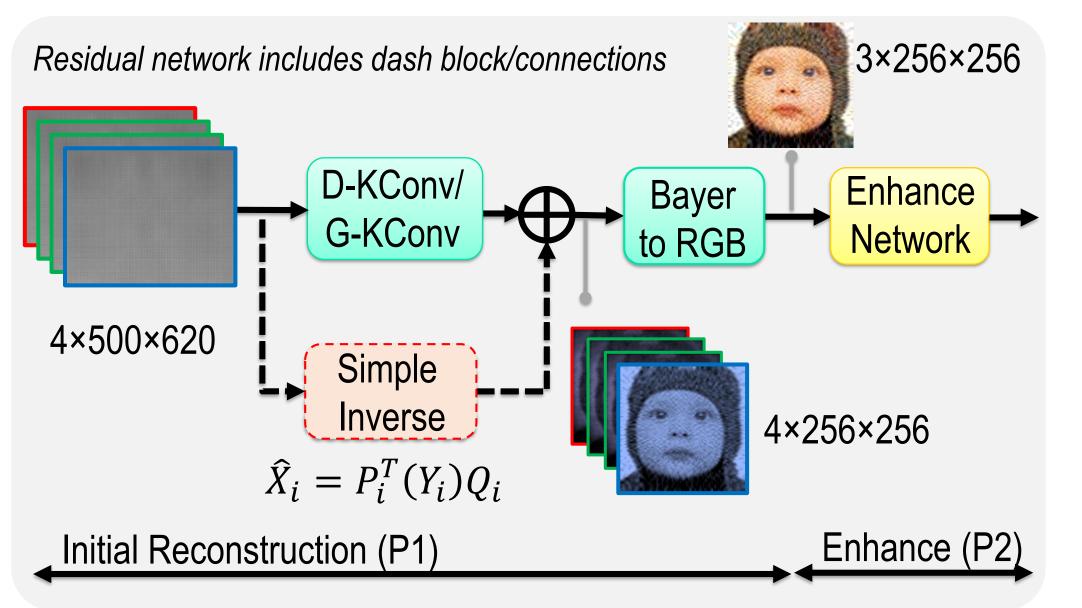
## 2. Deep FlatCam Reconstruction

Initial reconstruction with Kronecker Conv. (KConv)

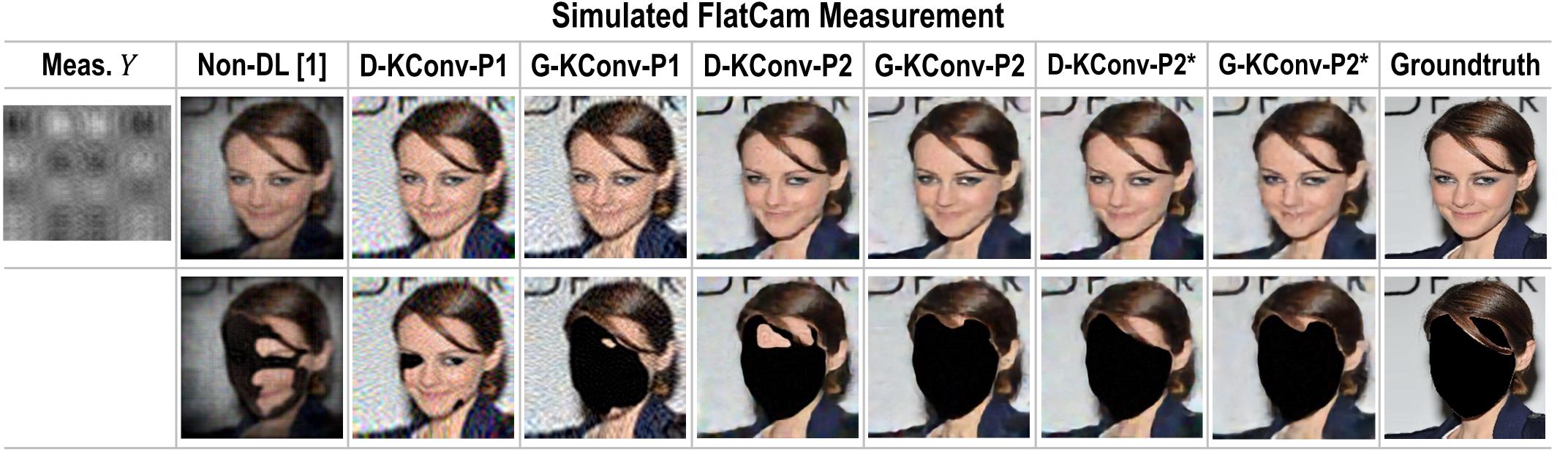




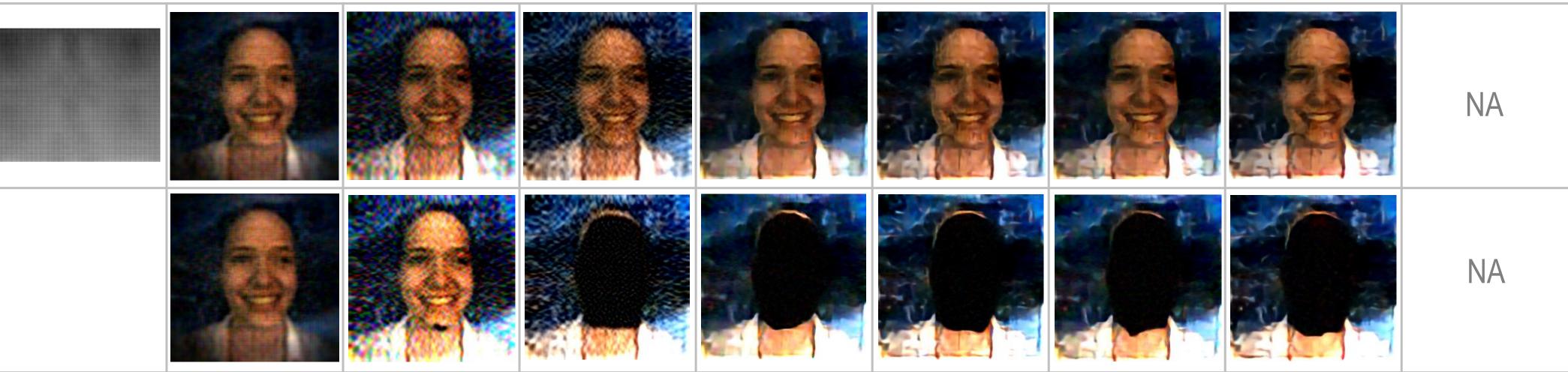
 Residual Reconstruction framework – two phases<sup>[2]</sup> Utilize the calibrated sampling matrices P, Q



## 5. Visual Quality of Reconstruction and Visual Privacy Protection



#### Real FlatCam Measurement from [2]



First and third row: the reconstructed images -  $\hat{X}$ ; second and forth row: the privacy protected images -  $\hat{X}_p$ Residual networks are marked with (\*)

Related Work

[1] M. S. Asif et al. Thin, lensless cameras using coded aperture and computation. IEEE Trans. Computational Imaging (TCI), 2017 [2] J. Tan et al. Face detection and verification using lensless cameras. IEEE Transaction on Computational Imaging (TCI), June 2019

[3] T. N. Canh et al. Multi-scale deep compressive sensing network. IEEE Inter. Conf. Visual Comm. Image Processing (VCIP), Dec. 2018.

### 3. Visual Privacy Protection

- Privacy protection in spatial domain  $(\widehat{X}_p)$  $\hat{X} = \hat{X}_p + \hat{X}_S$  s.t.  $\hat{X}_S = S \odot \hat{X}$
- Privacy protection in FlatCam meas. domain  $(Y_p)$  $Y_p \cong P \cdot \hat{X}_p \cdot Q^T = P \cdot (\hat{X} - \hat{X}_S) \cdot Q^T$ 
  - $= Y P \cdot \hat{X}_{p} \cdot Q^{T} = Y Y_{S}$
- Unify recovery & segmentation for visual privacy
  - Input/(output) is visually/(visually & privacy) protected Assume attacker cannot extract intermediate features
- → Resilient to transmission and encoder side attacks

## 4. Performance Evaluation

Reconstruction performance – PSNR (dB)

	Network	Initial Recons.		Enhance	
		D-KConv	G-KConv	D-KConv	G-KConv
SatE	Standard	21.83	22.53	24.01	24.09
Set5	Residual	22.53	22.63	24.17	24.04
Co+1 /	Standard	21.37	21.95	23.44	23.46
Set14	Residual	21.36	22.05	23.55	24.41

FlatCam<sup>[1]</sup>: Set 5 – 14.43 dB, Set14 – 13.89 dB

 Evaluate the visual privacy protection quality via Facial Pixel Accuracy (FPA) & Intersection over Union (IoU)

		Network	Initial Recons.		Enhance	
			D-KConv	G-KConv	D-KConv	G-KConv
FP	ΕDΛ	Standard	0.4090	0.9417	0.9386	0.9631
	ГРА	Residual	0.4082	0.9427	0.9608	0.9646
lol	ااما	Standard	0.3854	0.8521	0.8483	0.8678
	100	Residual	0.3847	0.8527	0.8678	0.8766

Org/FlatCam<sup>[1]</sup>: FPA – 0.9340/0.6429, IoU – 0.8765/0.8765

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