

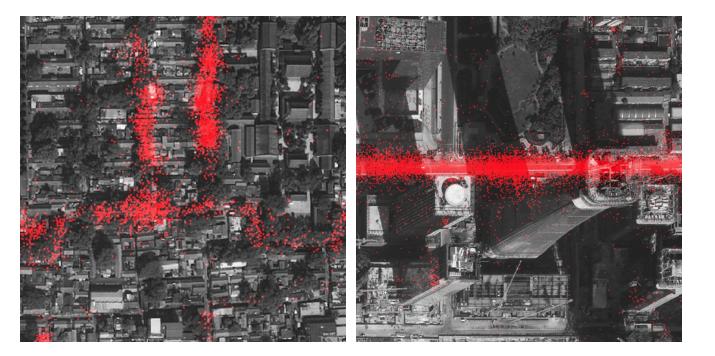


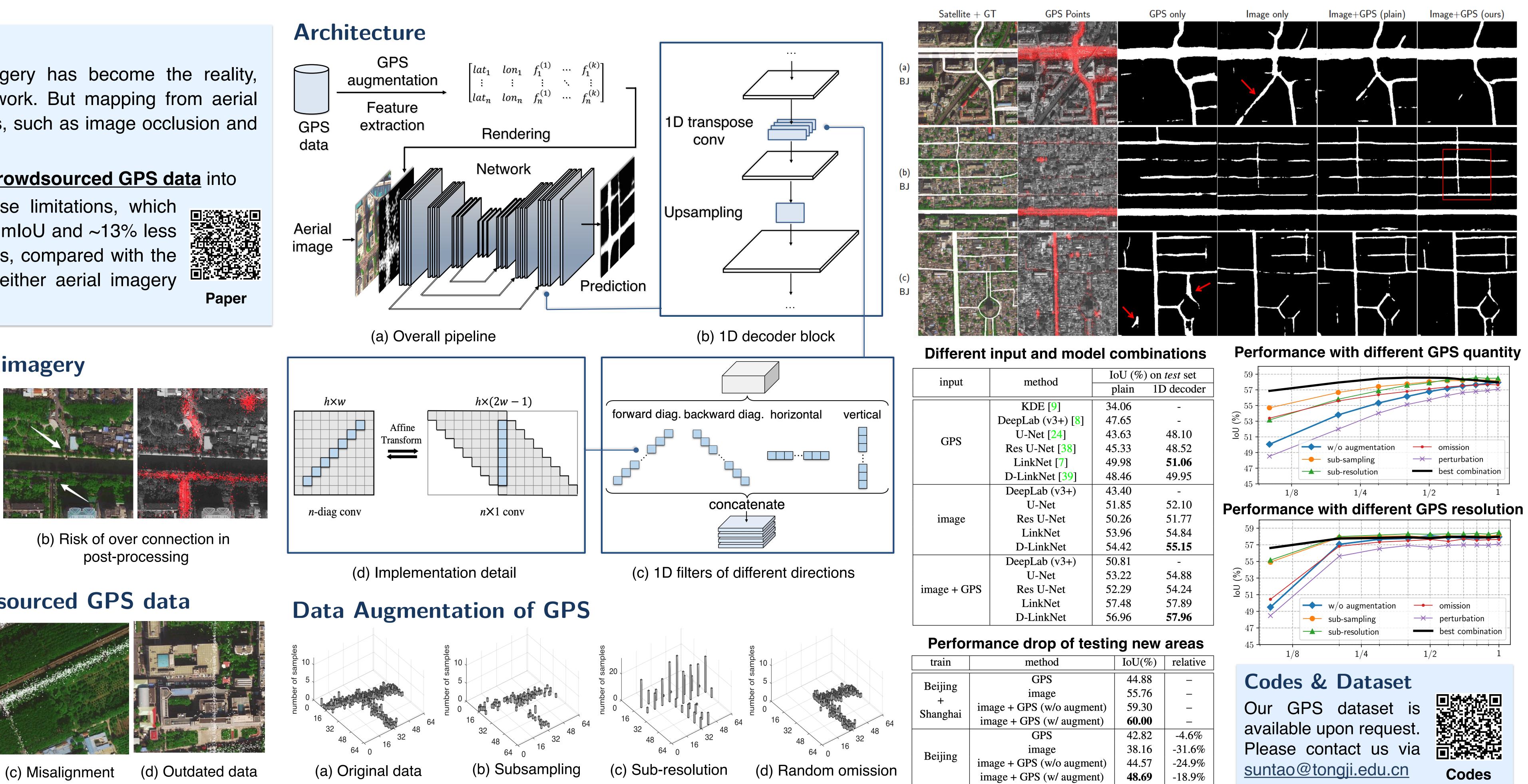
## Introduction

Al mapping from satellite imagery has become the reality, replacing the tedious manual work. But mapping from aerial imagery alone has its limitations, such as image occlusion and connectivity issues.

Here we propose to **integrate crowdsourced GPS data** into CNN models to overcome these limitations, which shows at least ~5% boosting in mIoU and ~13% less dropping in predicting new areas, compared with the state-of-the-art methods using either aerial imagery or GPS data alone.

## Typical issues with aerial imagery





(a) Occlusion by trees, buildings and their shadows

## **Typical issues with crowdsourced GPS data**



(a) Excessive noise



(b) Waiting area



# Leveraging Crowdsourced GPS Data for Road Extraction from Aerial Imagery Tao Sun, Zonglin Di, Pengyu Che, Chun Liu and Yin Wang\*, Tongji University

input	method	IoU (%) on <i>test</i> set	
input		plain	1D decoder
GPS	KDE [9]	34.06	-
	DeepLab (v3+) [8]	47.65	-
	U-Net [24]	43.63	48.10
OF2	Res U-Net [38]	45.33	48.52
	LinkNet [7]	49.98	51.06
	D-LinkNet [39]	48.46	49.95
image	DeepLab (v3+)	43.40	-
	U-Net	51.85	52.10
	Res U-Net	50.26	51.77
	LinkNet	53.96	54.84
	D-LinkNet	34.06   [8] 47.65   43.63   8] 45.33   49.98   9] 48.46   +) 43.40   51.85   50.26   53.96   54.42	55.15
image + GPS	DeepLab (v3+)	50.81	-
	U-Net	53.22	54.88
	Res U-Net	52.29	54.24
	LinkNet	57.48	57.89
	D-LinkNet	56.96	57.96

train	method	IoU(%)	relative
Beijing + Shanghai	GPS	44.88	_
	image	55.76	_
	image + GPS (w/o augment)	59.30	_
	image + GPS (w/ augment)	60.00	_
Beijing	GPS	42.82	-4.6%
	image	38.16	-31.6%
	image + GPS (w/o augment)	44.57	-24.9%
	image + GPS (w/ augment)	48.69	-18.9%





### LONG BEACH CALIFORNIA June 16-20, 2019