

ETH zürich¹



EUROPEAN CONFERENCE ON COMPUTER VISION



GeoCalib

Learning Single-image Calibration
with Geometric Optimization



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Veicht¹



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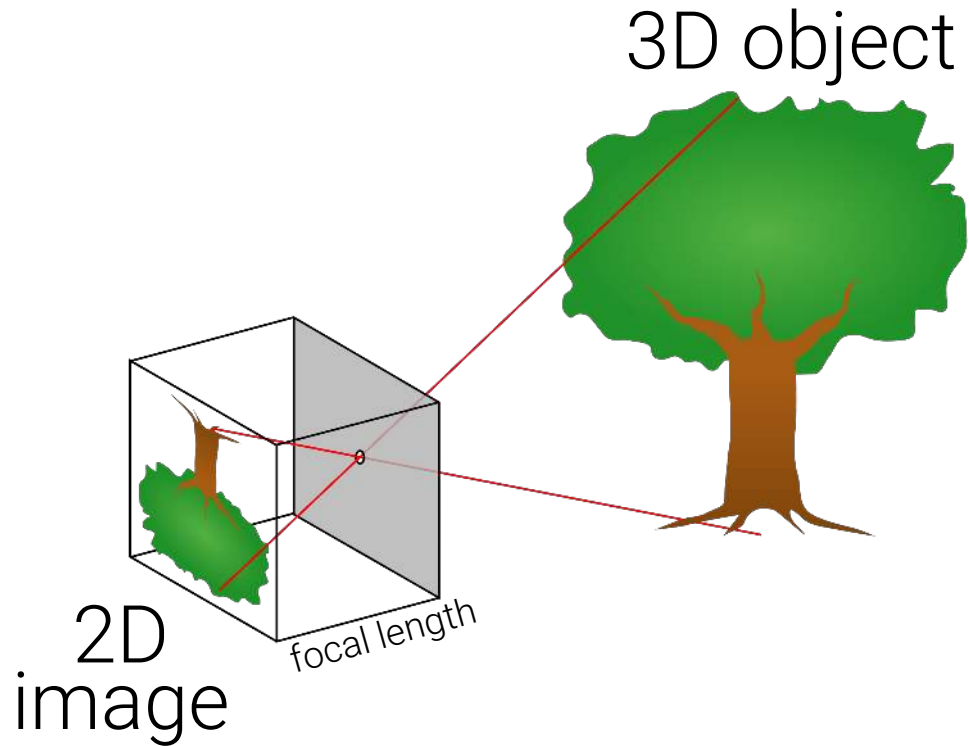
Philipp
Lindenberger¹



Marc
Pollefeys^{1,2}

github.com/cvg/GeoCalib

Camera Calibration



focal length

intrinsic parameters

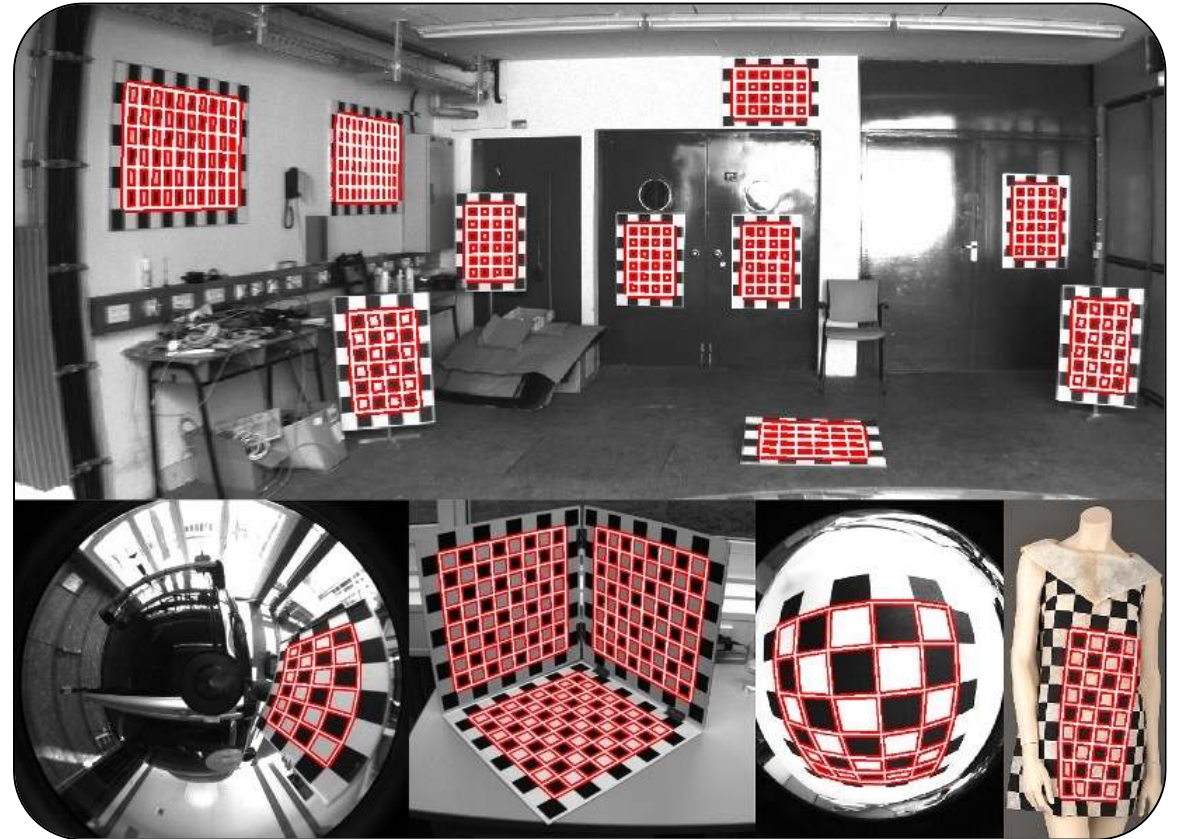
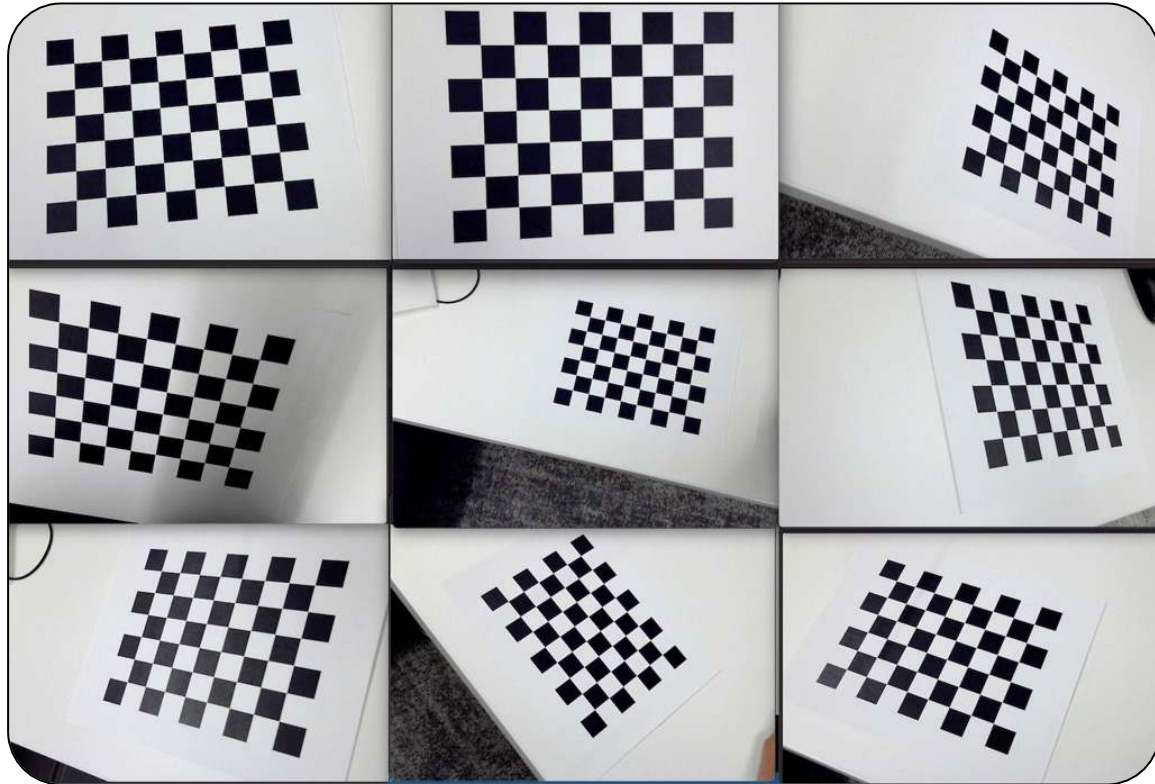
lens distortion



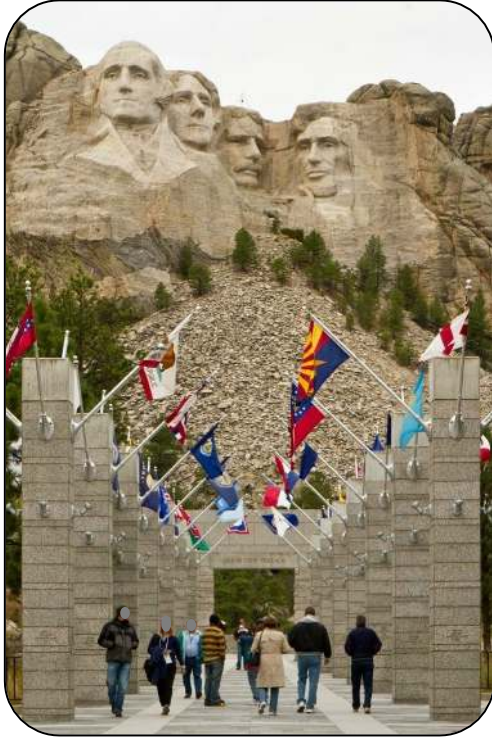


Controlled Conditions

Kalibr

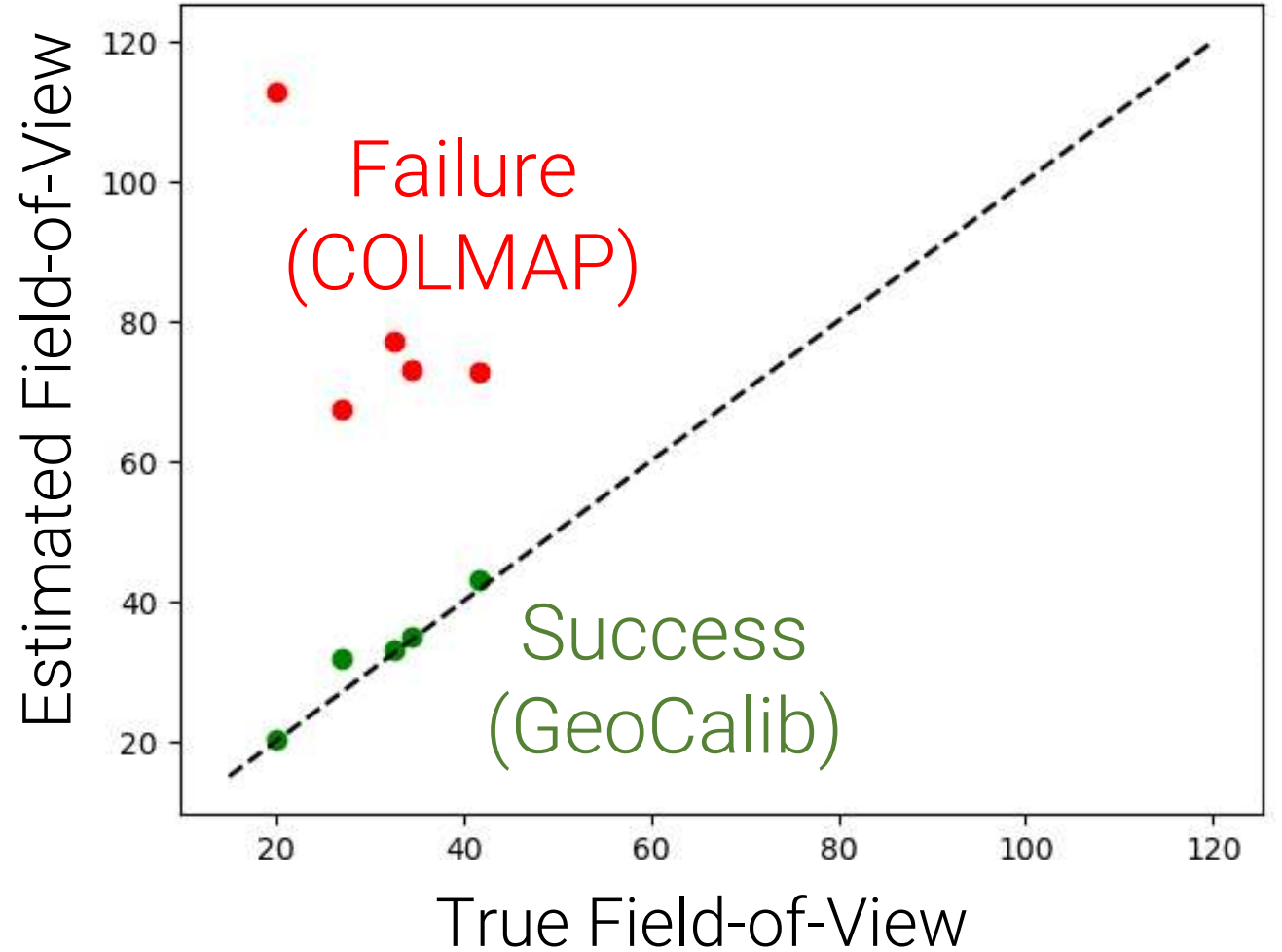


Uncontrolled Conditions



Photogrammetry
Structure-from-Motion

Uncontrolled Conditions



Camera Calibration



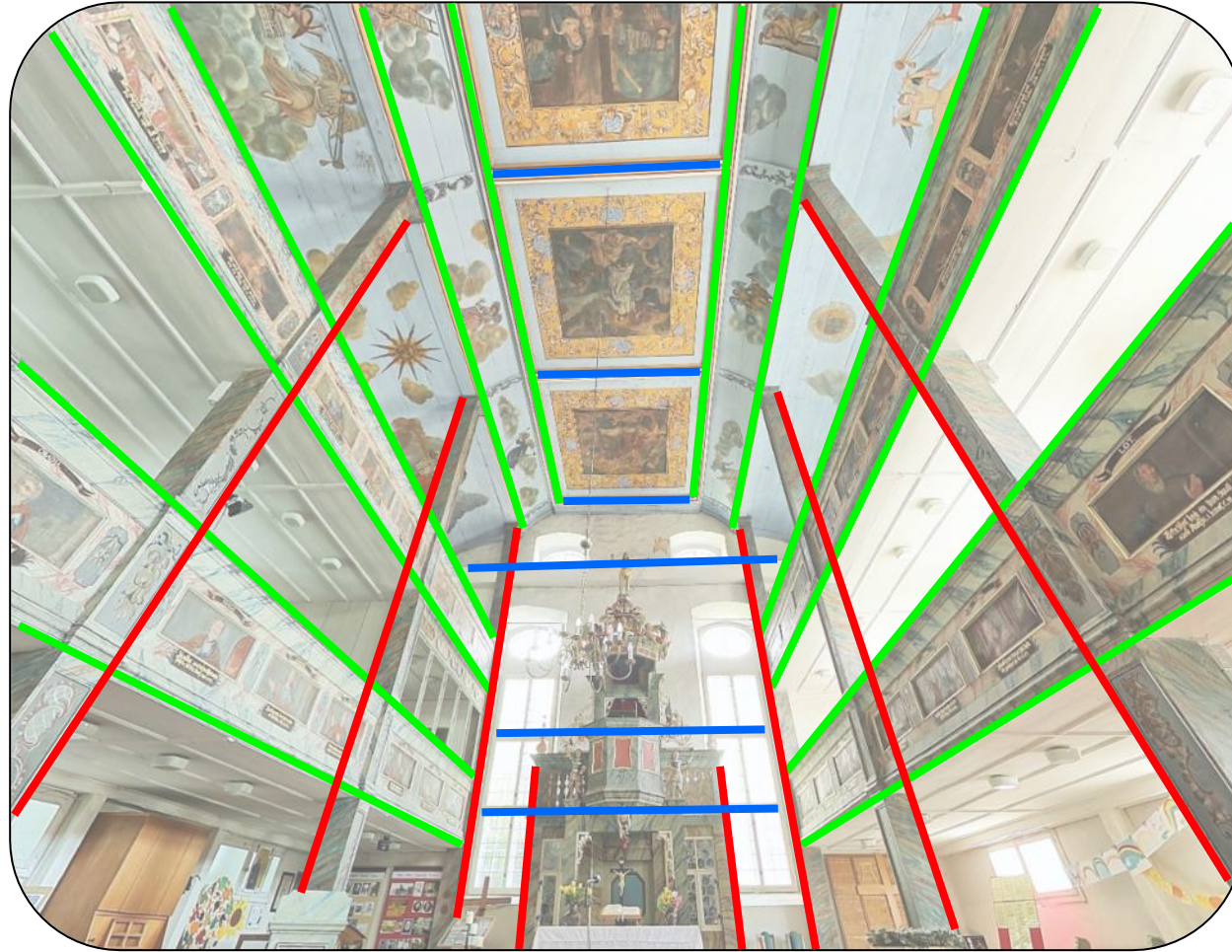
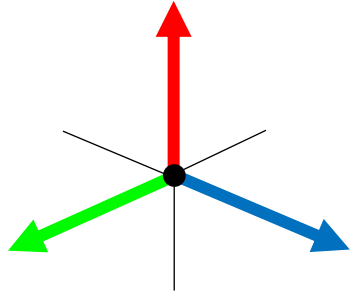
roll



pitch

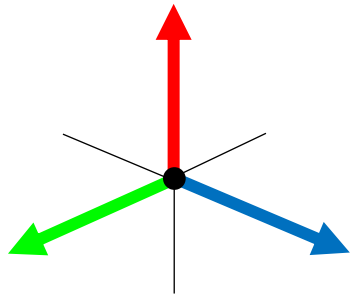
extrinsic parameters = gravity direction

Single-image Camera Calibration



gravity & intrinsic parameters

Existing algorithms



GT and
estimated
horizon lines

many lines → accurate

Existing algorithms



many lines → **accurate**



few lines → **failure**

Deep Learning

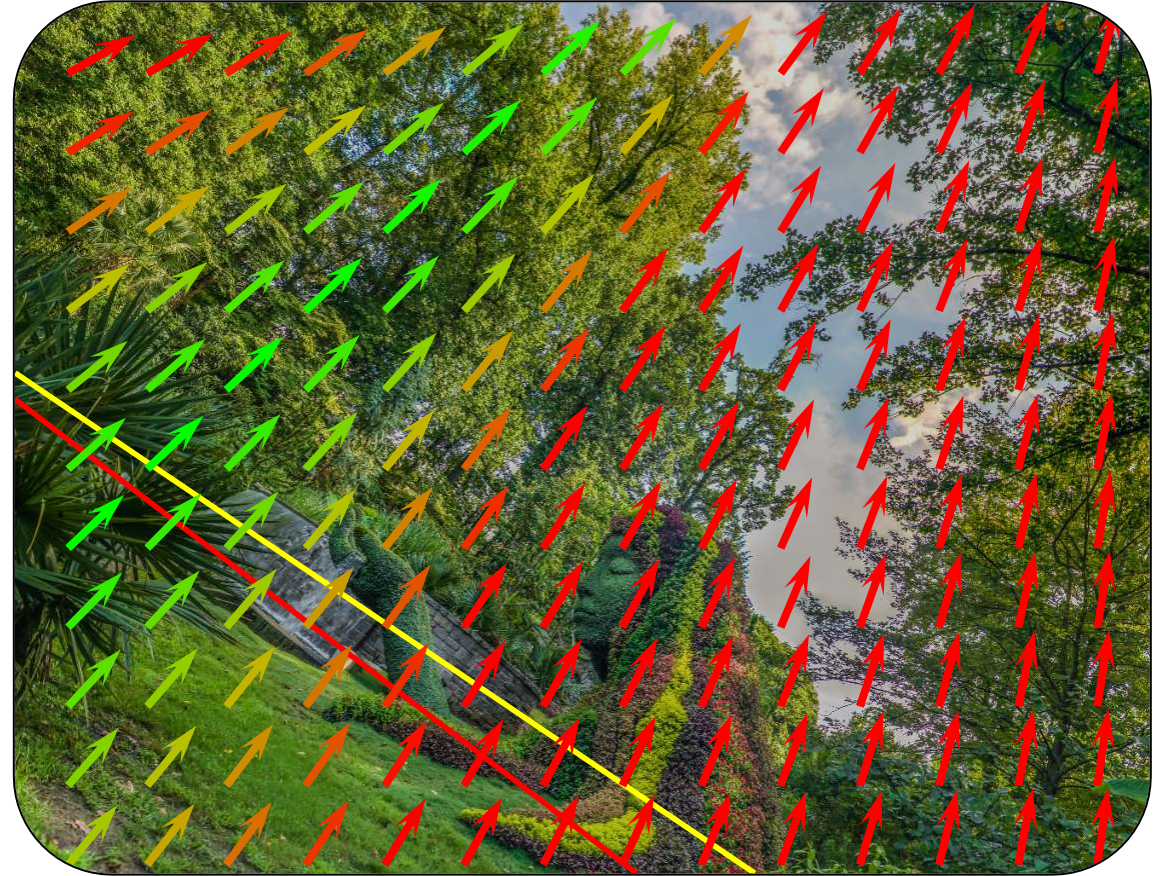
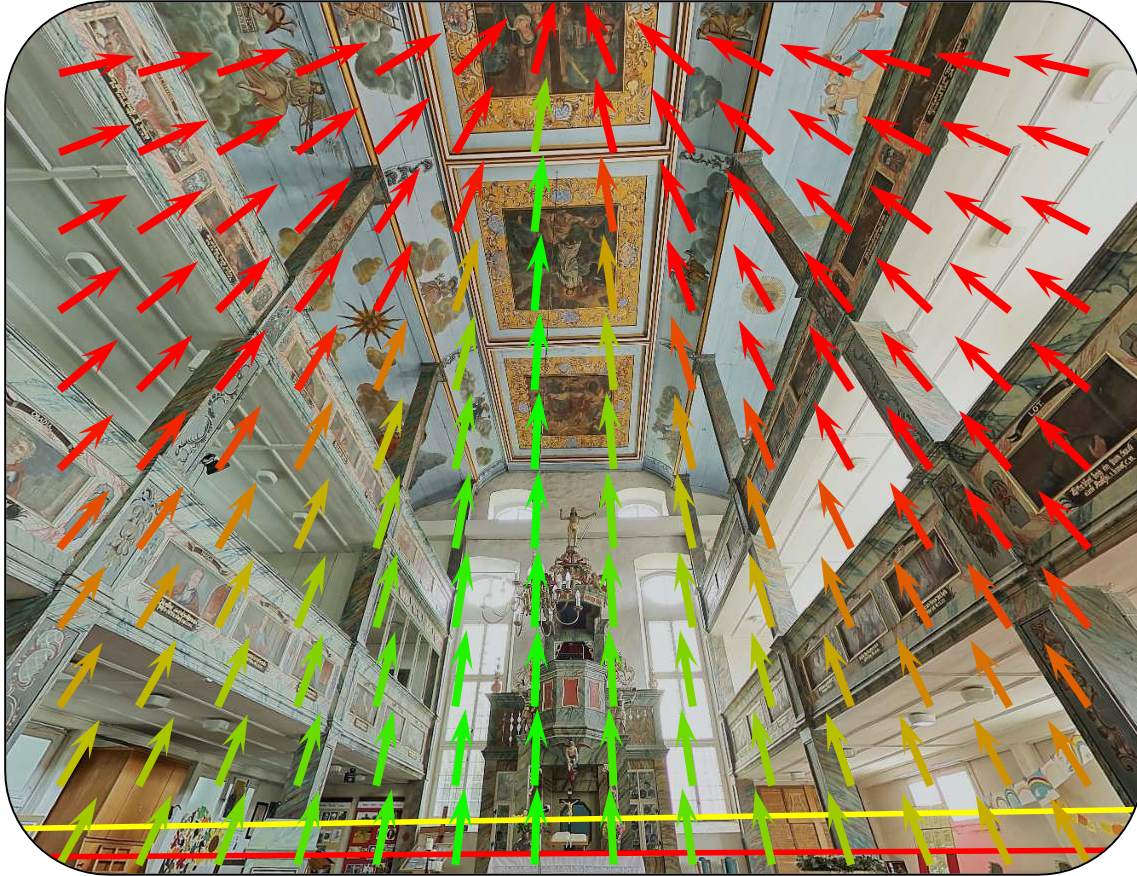


black-box
Deep Neural
Network



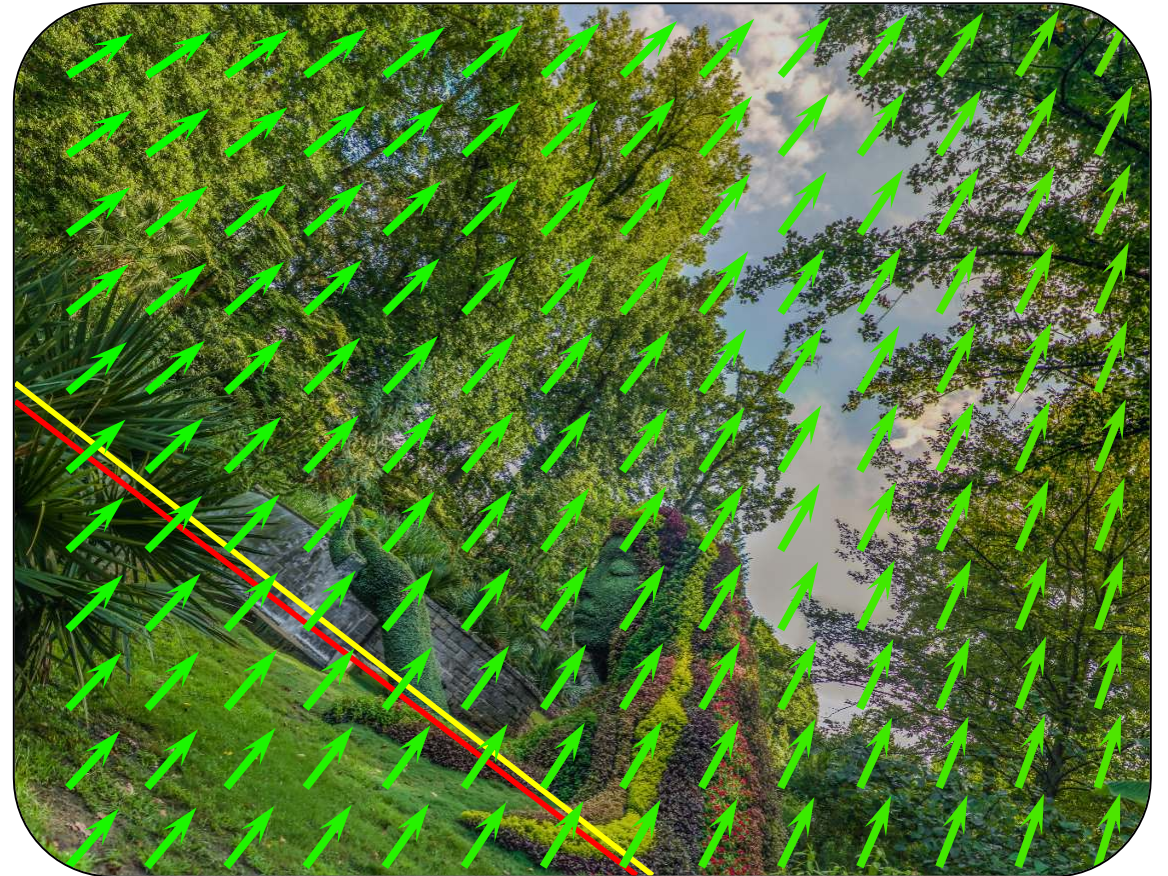
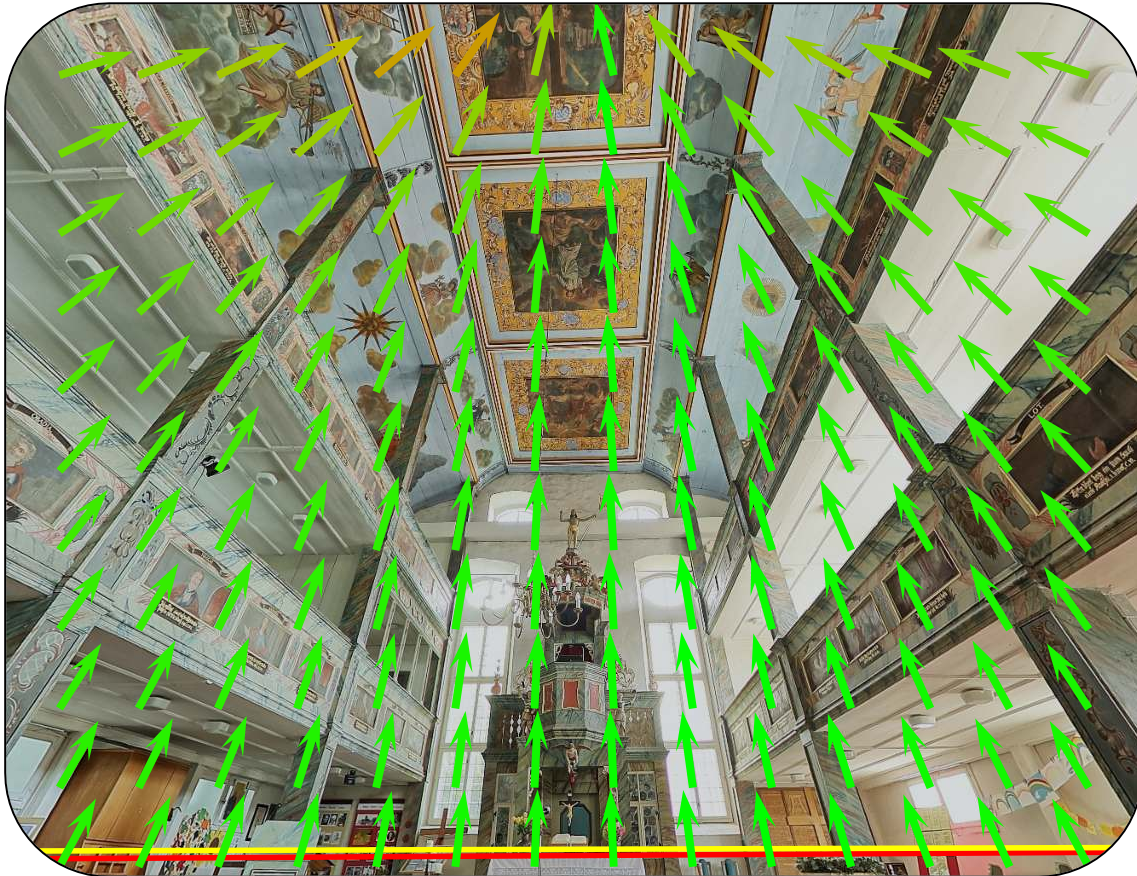
gravity
& intrinsic
parameters

Deep Learning



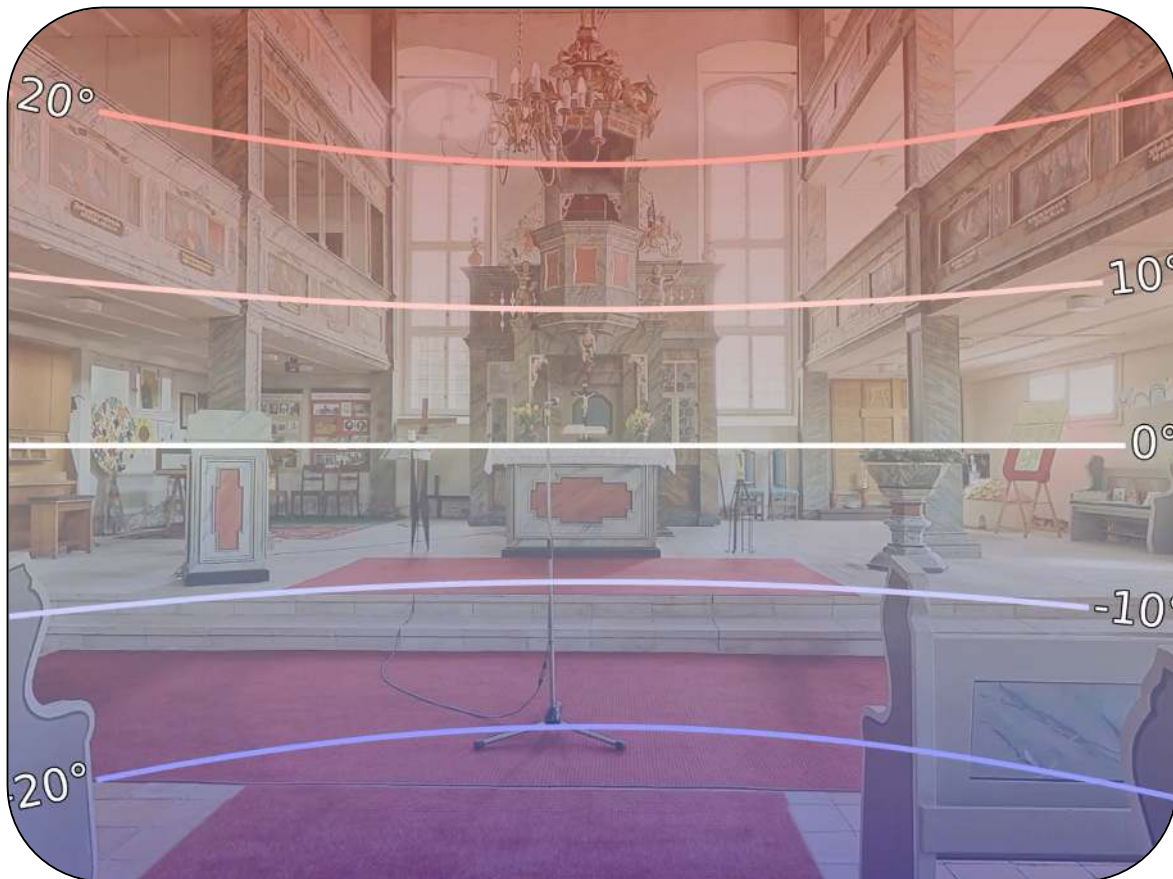
more robust but less accurate

GeoCalib = Learning + Geometry

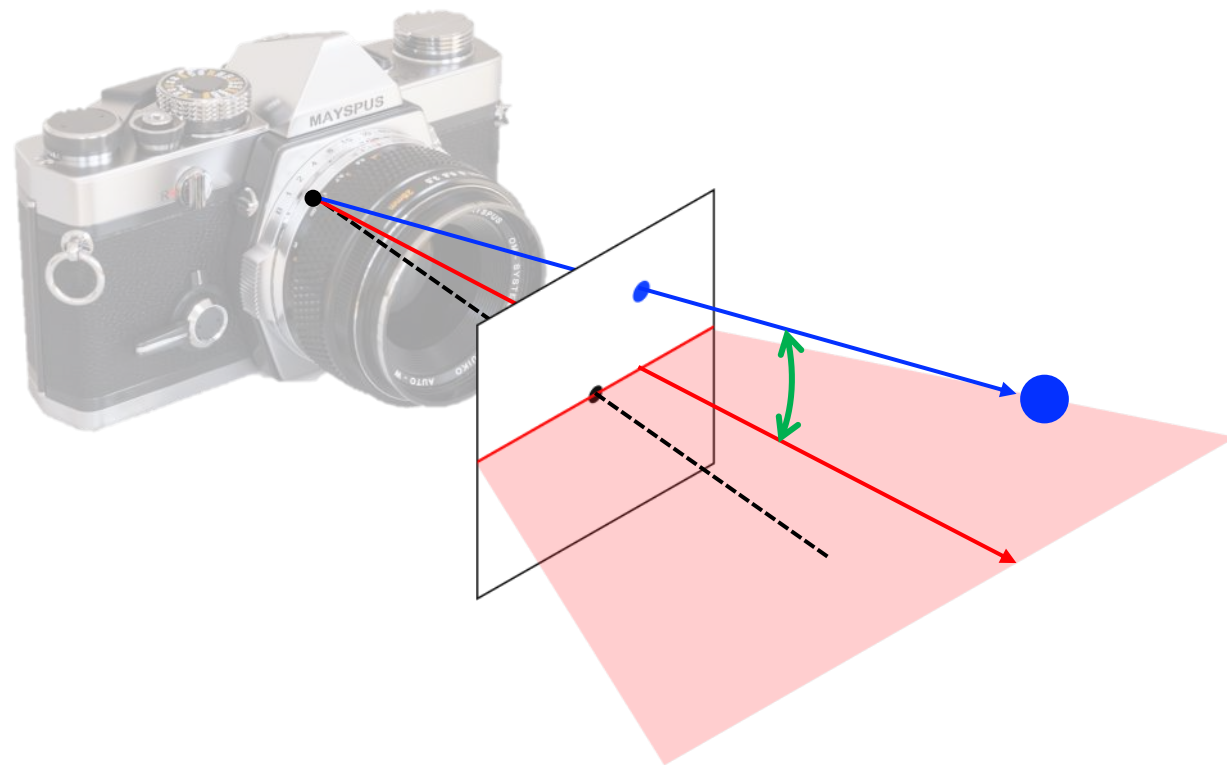


both more robust and more accurate

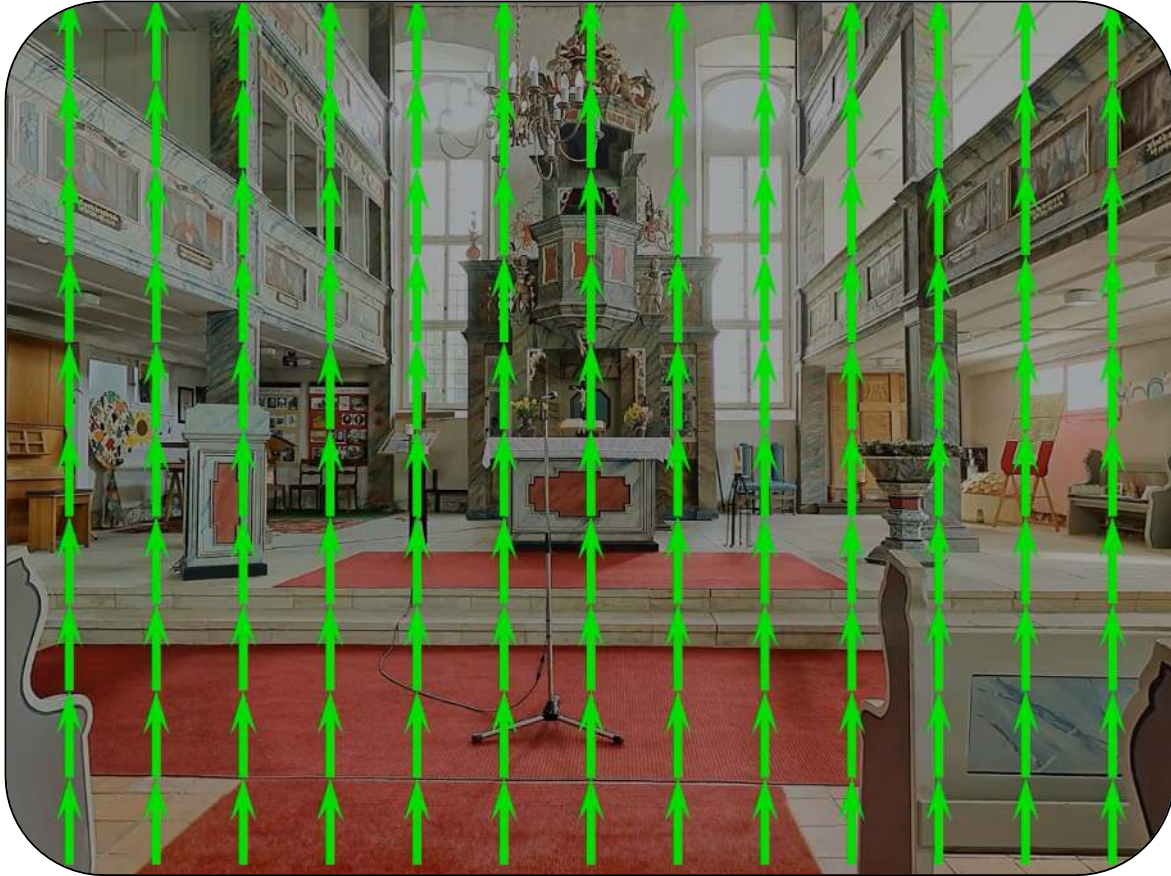
Perspective Field [Jin et al, CVPR 2023]



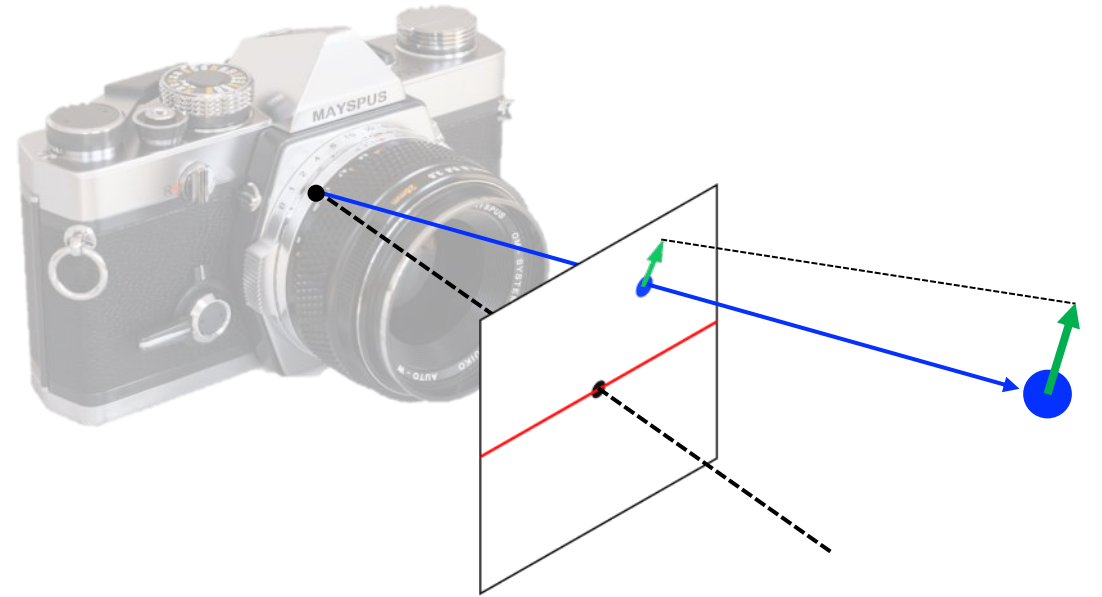
Latitude



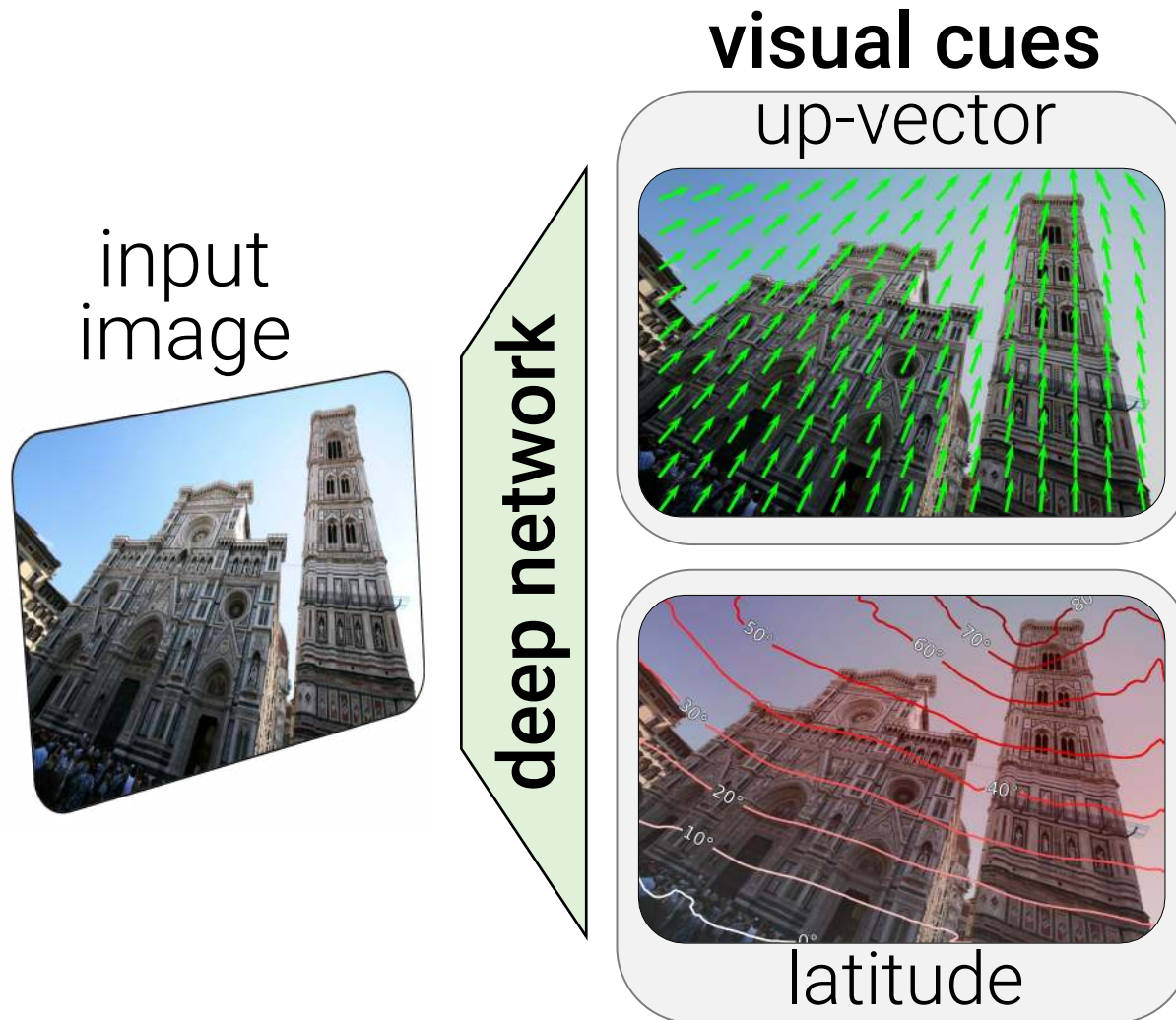
Perspective Field [Jin et al, CVPR 2023]



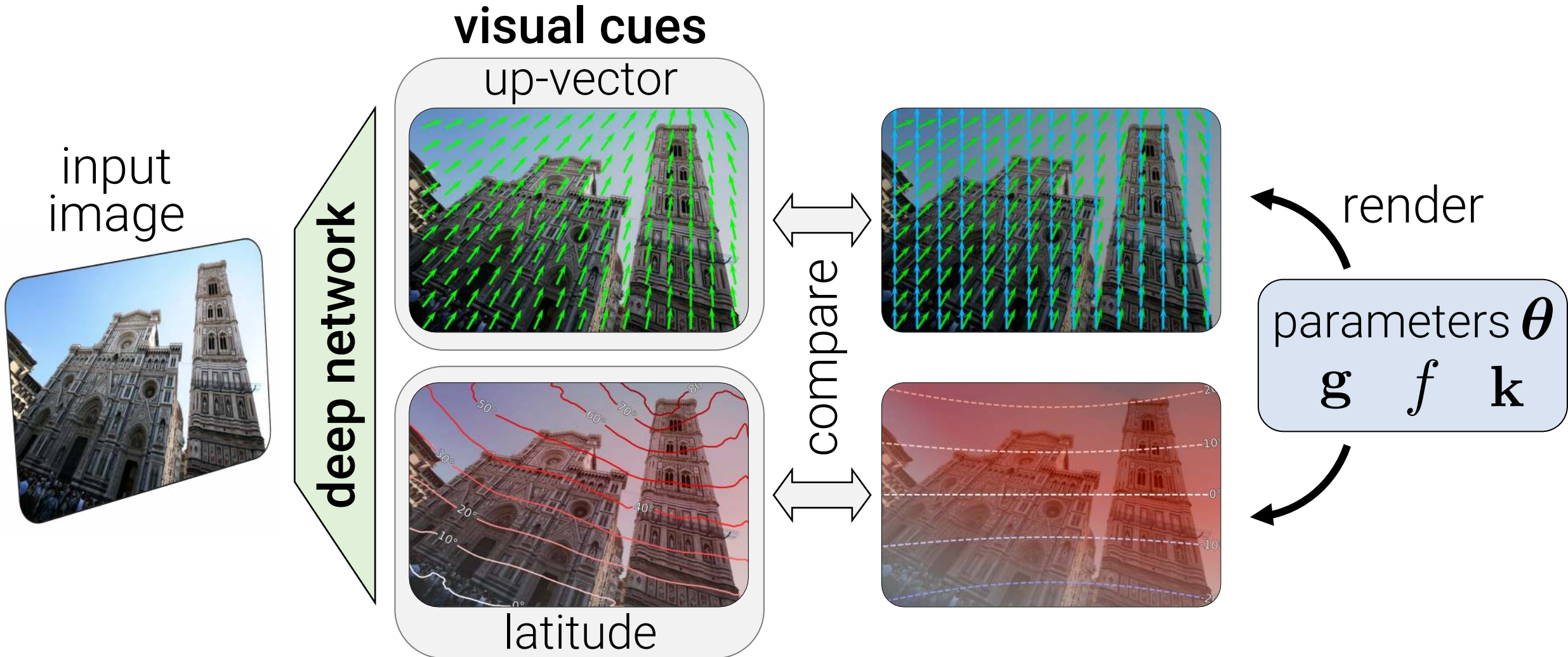
Up-vector



GeoCalib – Architecture



GeoCalib – Architecture



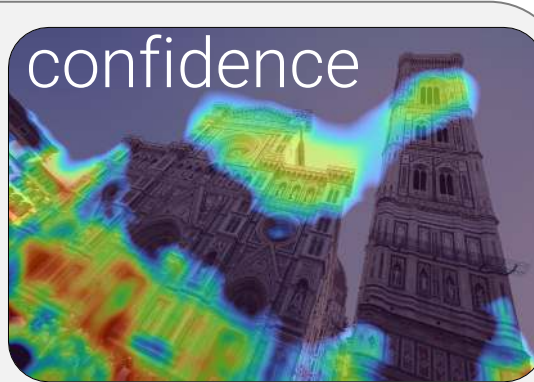
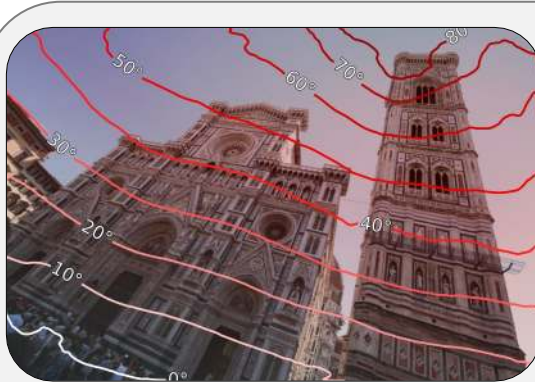
GeoCalib – Architecture

visual cues

up-vector



confidence



confidence

latitude

input
image



deep
network

compare



GeoCalib – Architecture

visual cues

up-vector



confidence

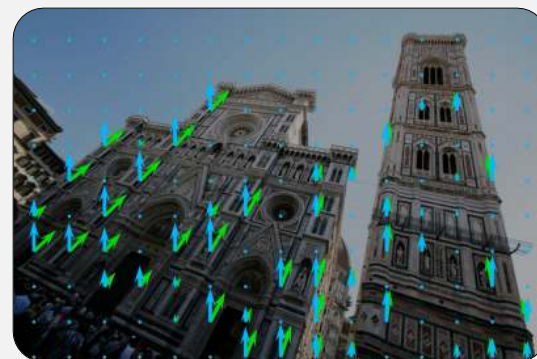


confidence



latitude

optimization



LM solver

parameters θ
 g f k

gravity + intrinsics

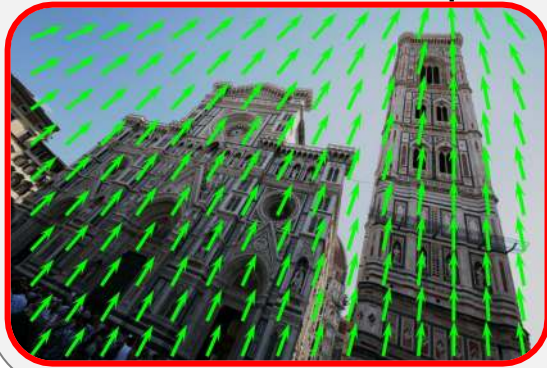
deep network



GeoCalib – Architecture

visual cues

up-vector



confidence



confidence



latitude

optimization



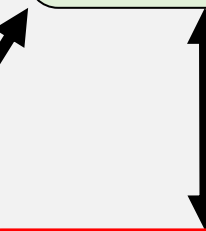
LM solver

parameters θ
 g f k

gravity + intrinsics

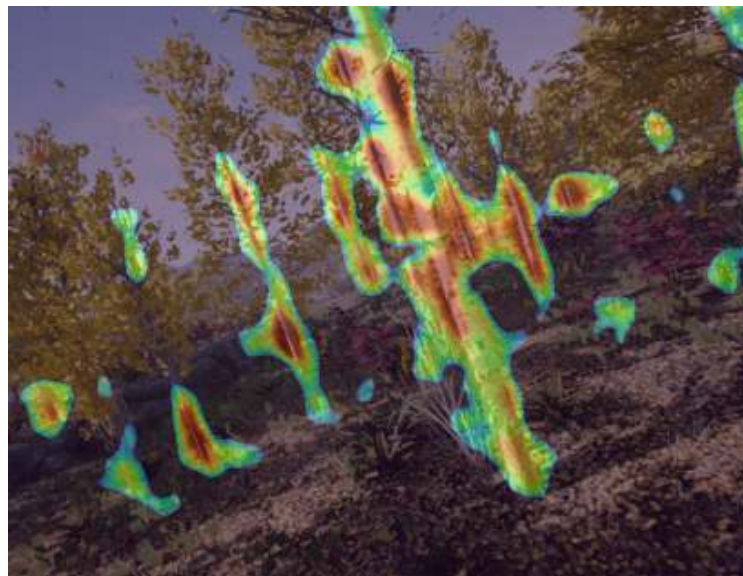
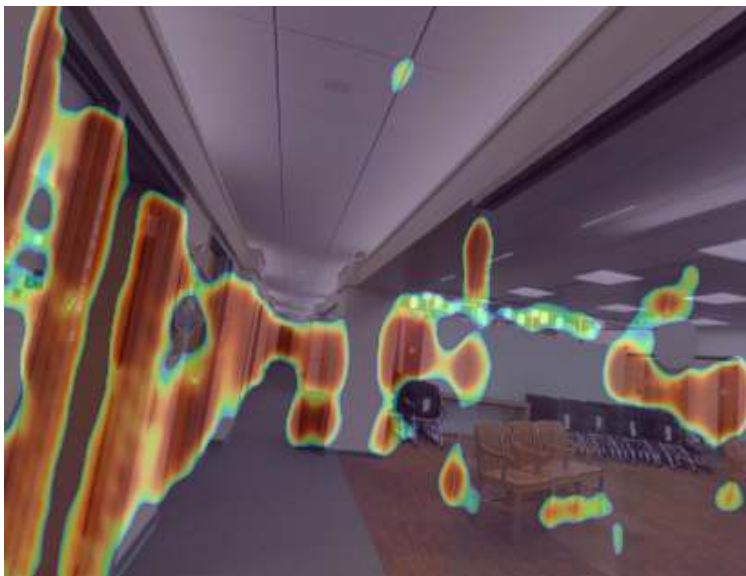
deep network

loss with GT

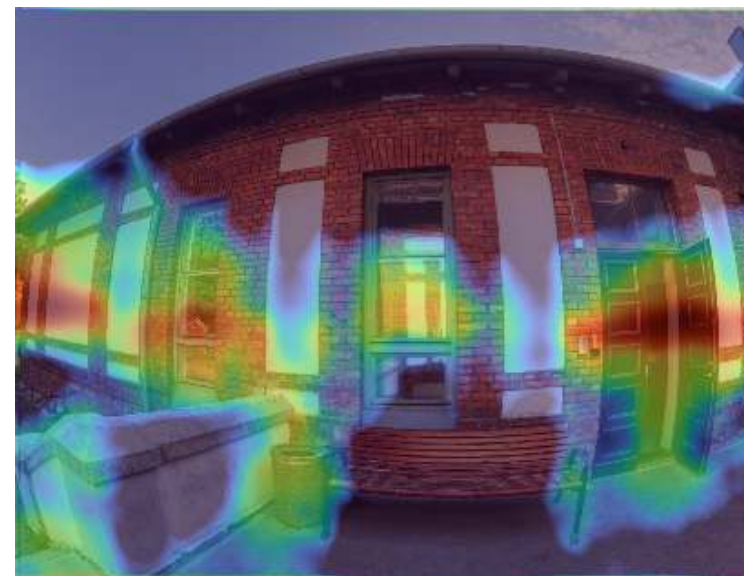
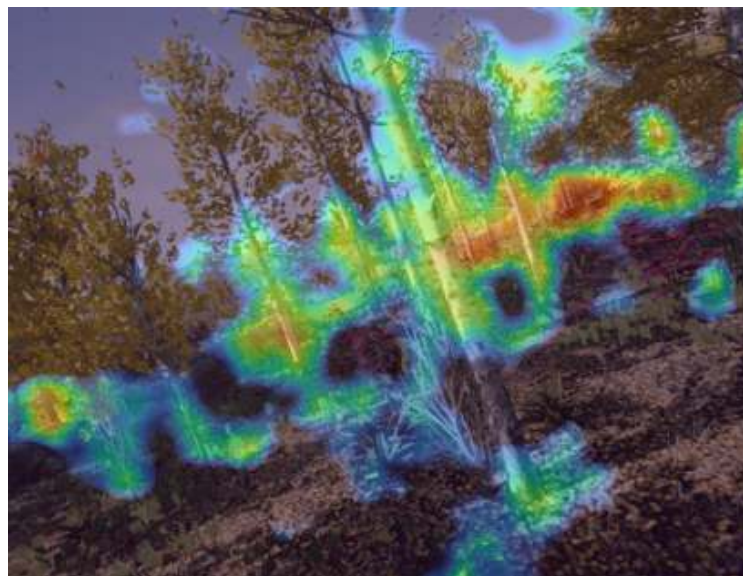
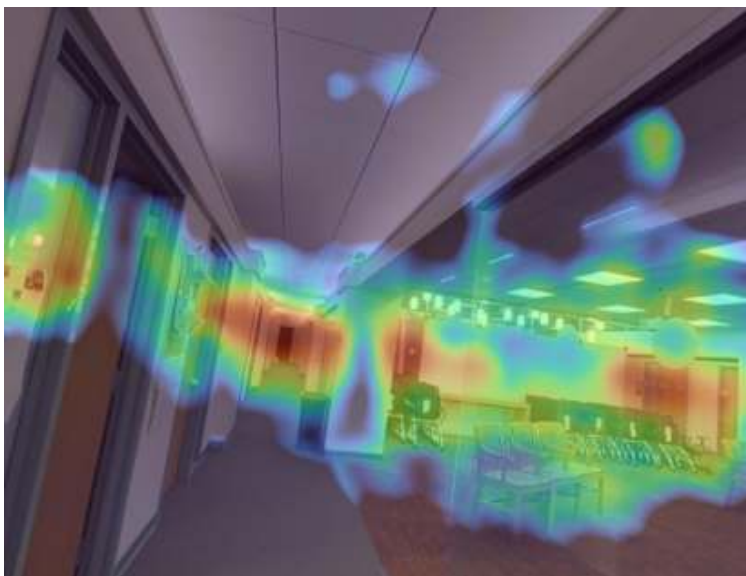


Learned confidences

Up-vector



Latitude

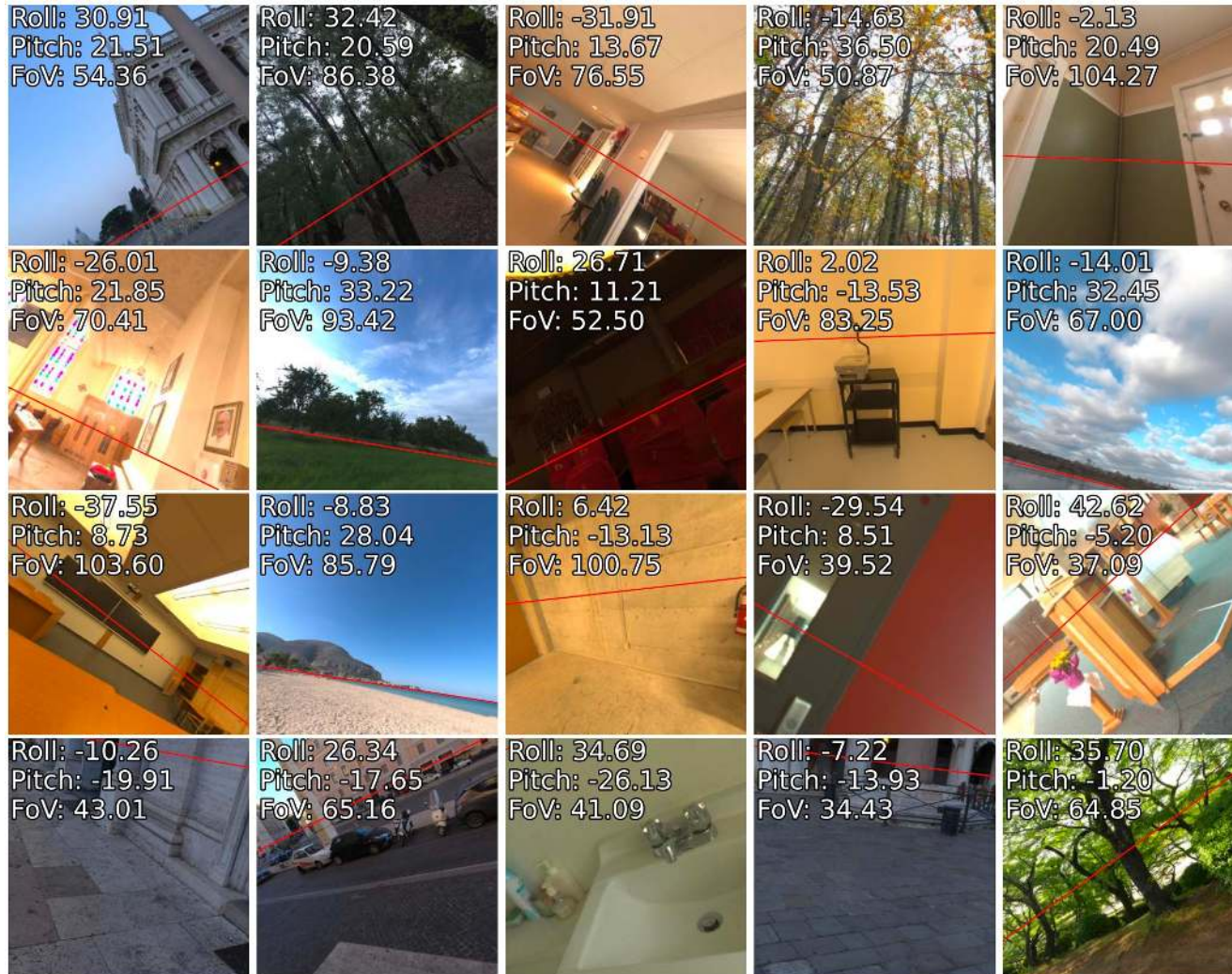


1



0

New dataset: OpenPano



Previous research is based on proprietary datasets

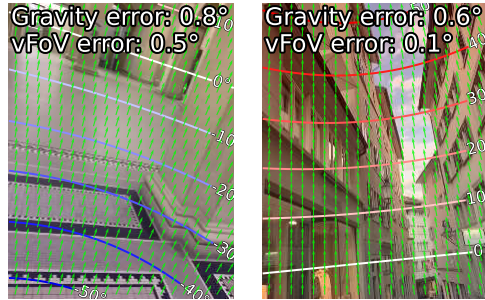
- SUN360, 360Cities

OpenPano: public sources

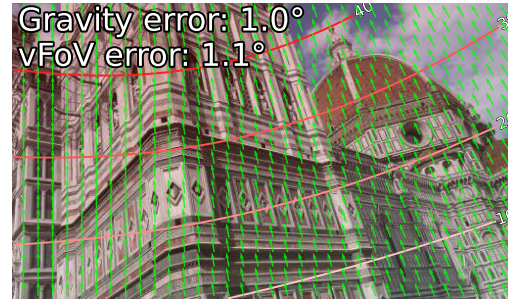
→ GeoCalib is fully reproducible

Evaluation

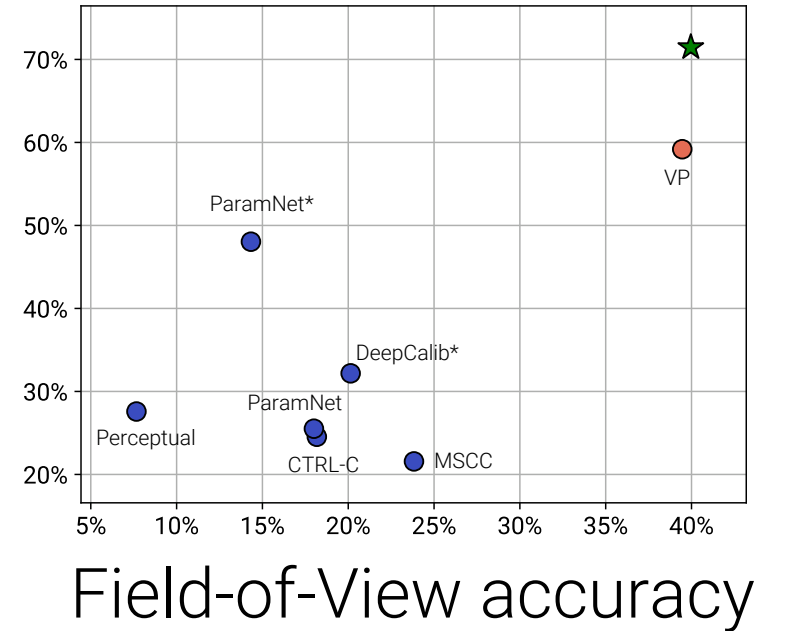
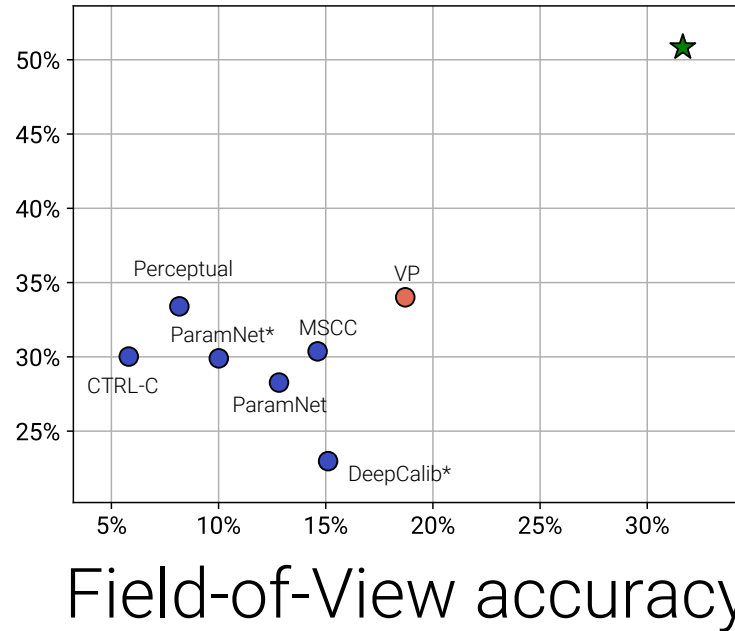
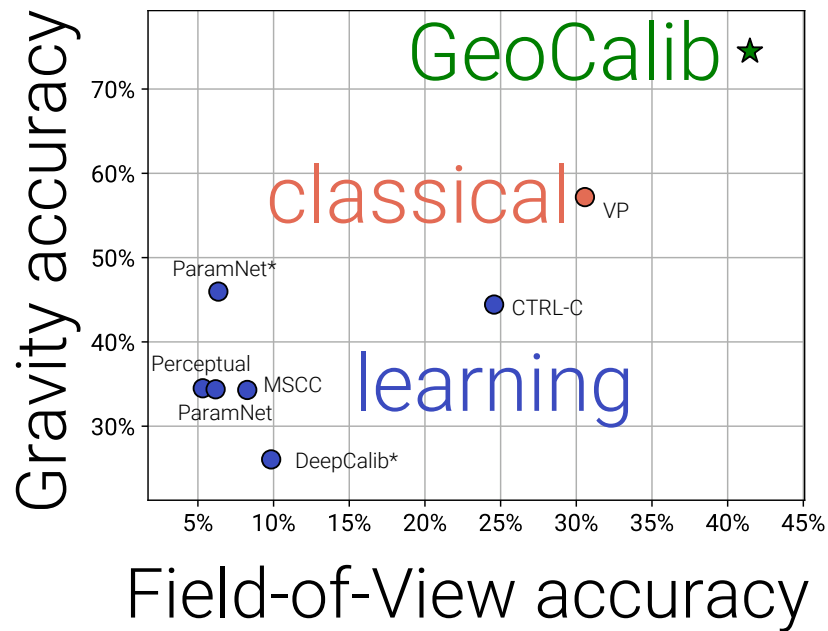
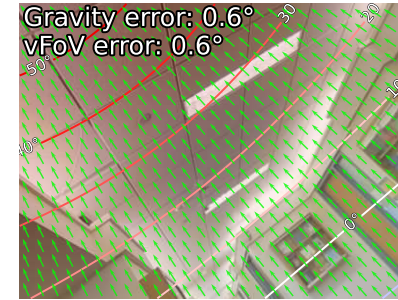
LaMAR - phones



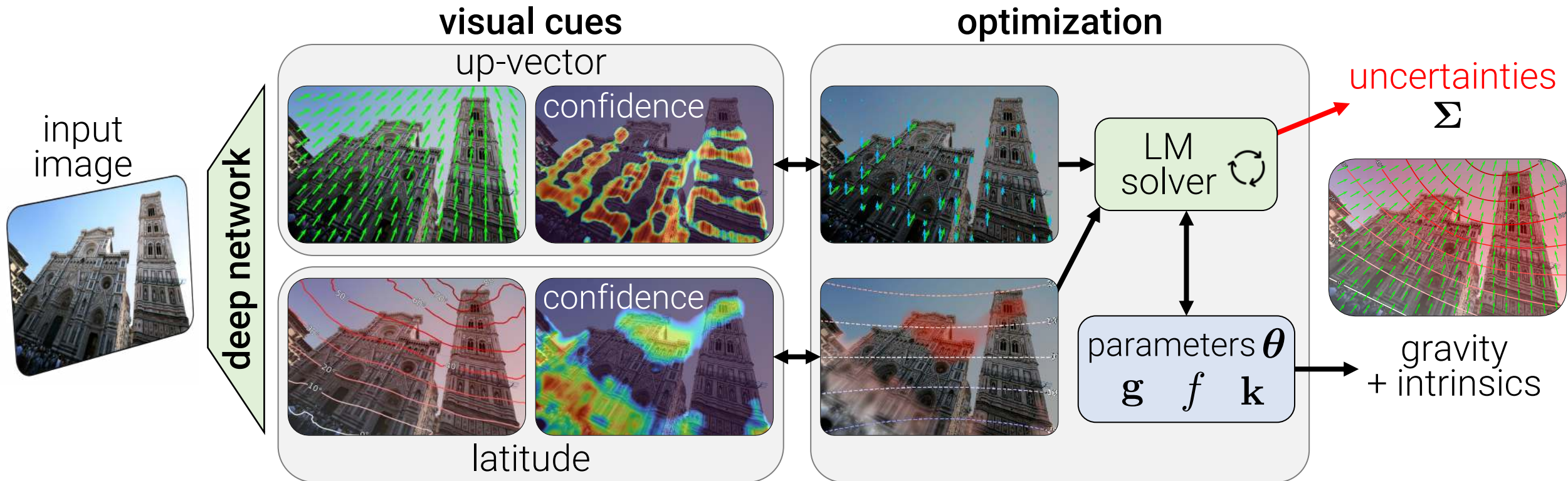
MegaDepth - outdoor



Stanford2D3D - indoor



Practical benefits



GeoCalib estimates reliable uncertainties

true gravity error vs predicted uncertainty

$0.5^\circ / 1.2^\circ$



$0.9^\circ / 1.4^\circ$



$2.1^\circ / 2.2^\circ$



$3.4^\circ / 3.2^\circ$



$3.4^\circ / 4.7^\circ$



$0.2^\circ / 0.7^\circ$



$0.7^\circ / 1.0^\circ$



$1.3^\circ / 1.5^\circ$



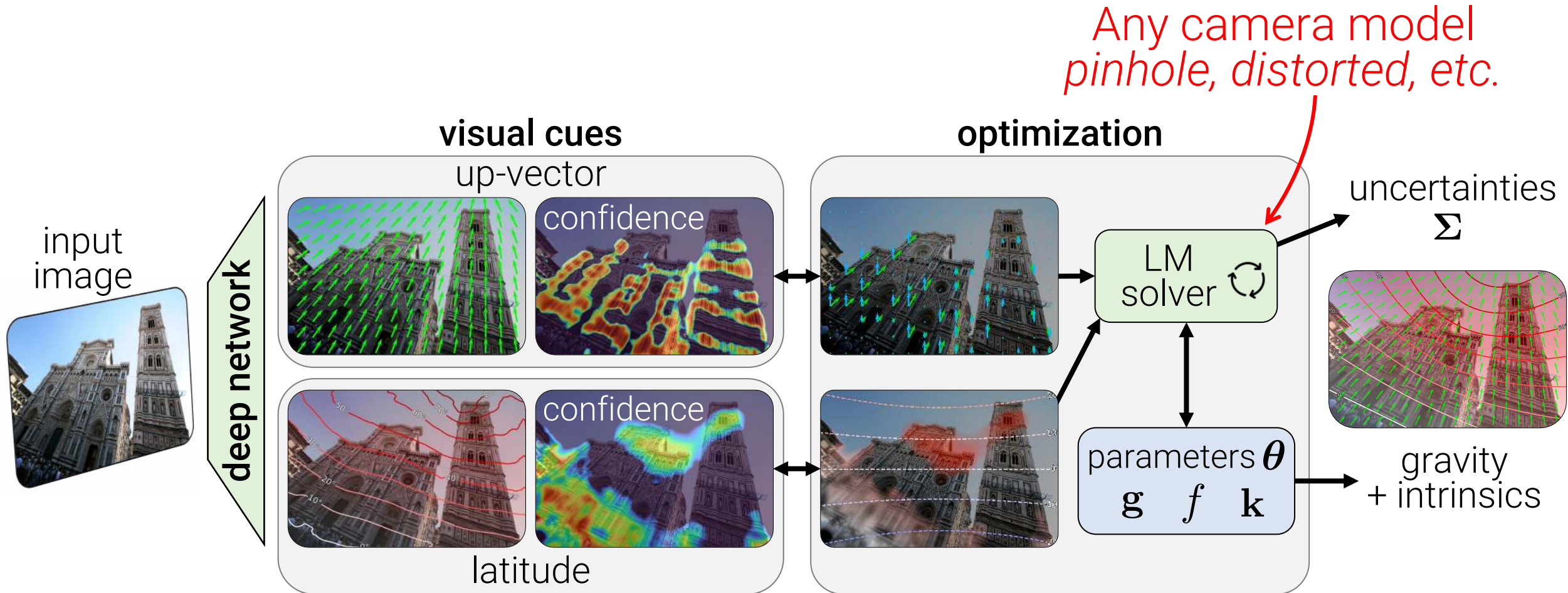
$2.4^\circ / 2.6^\circ$



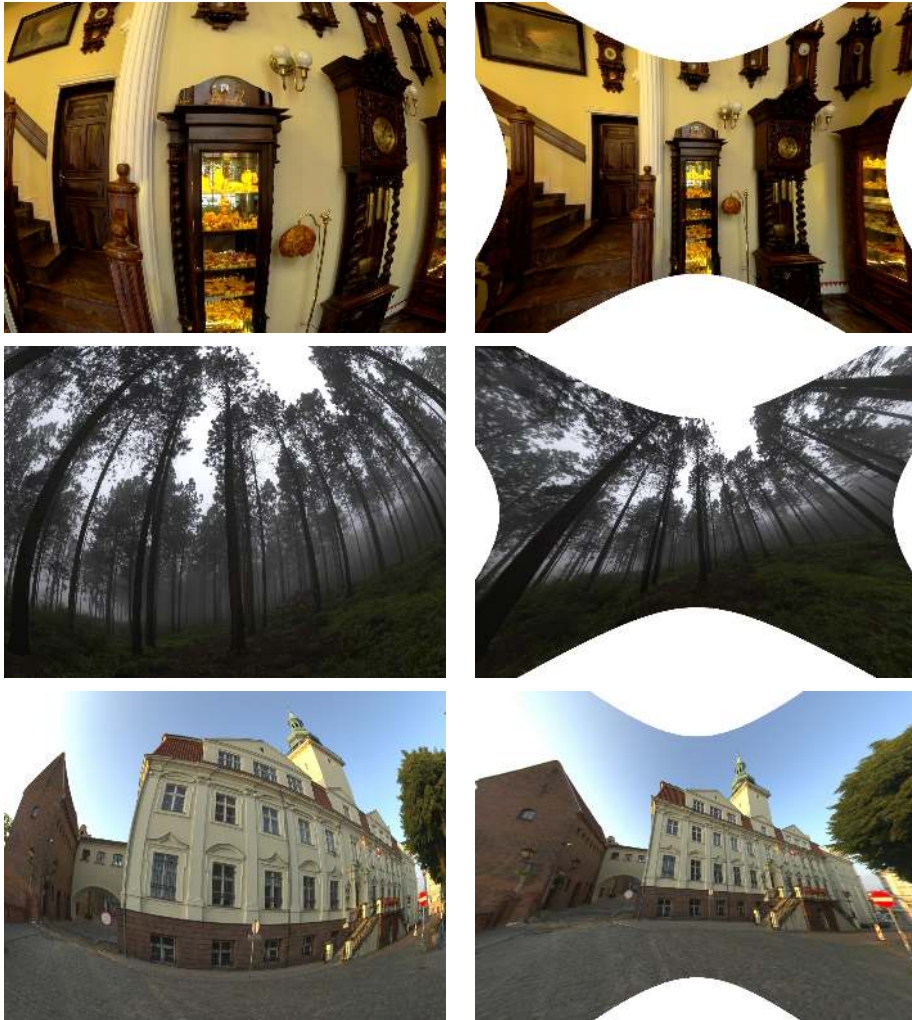
$4.3^\circ / 4.8^\circ$

more difficult examples 

Practical benefits

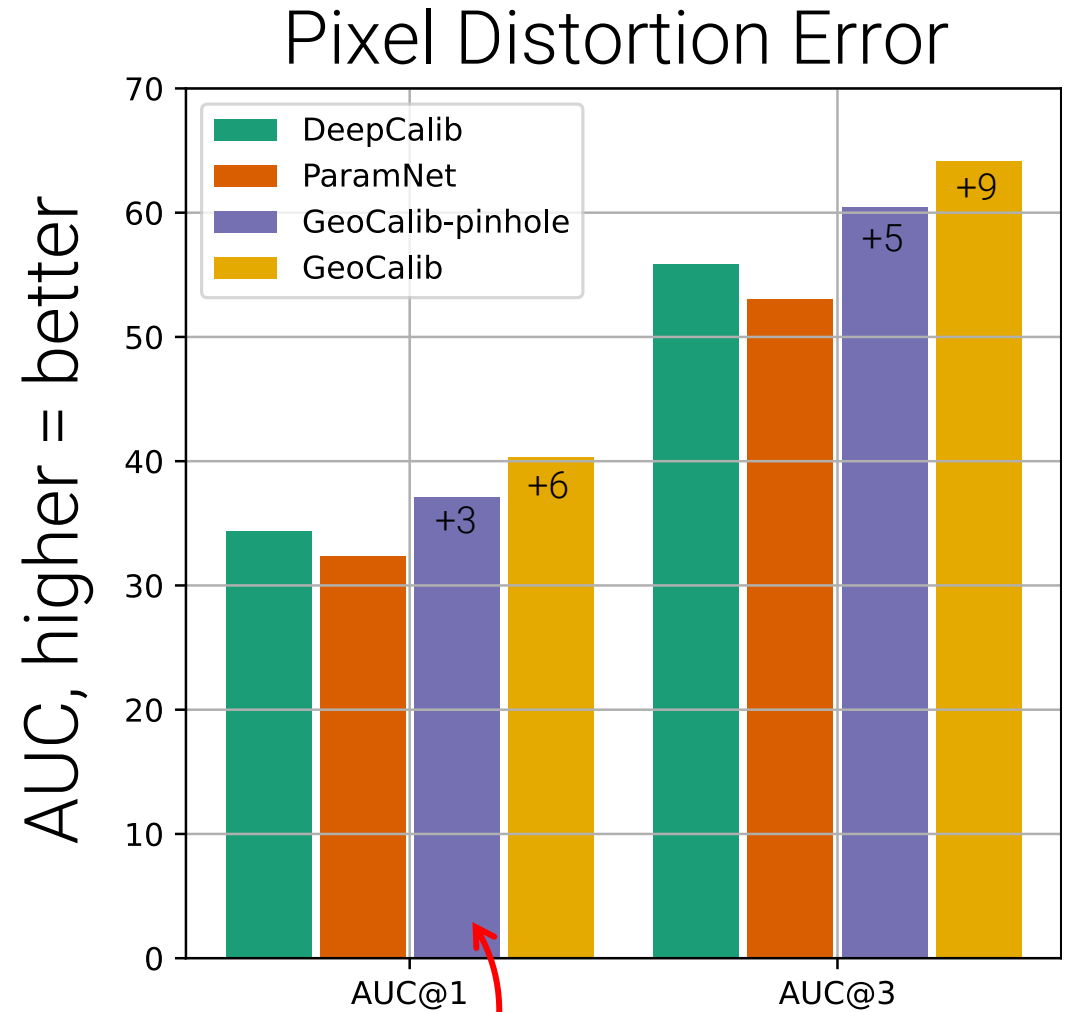


GeoCalib can handle lens distortion



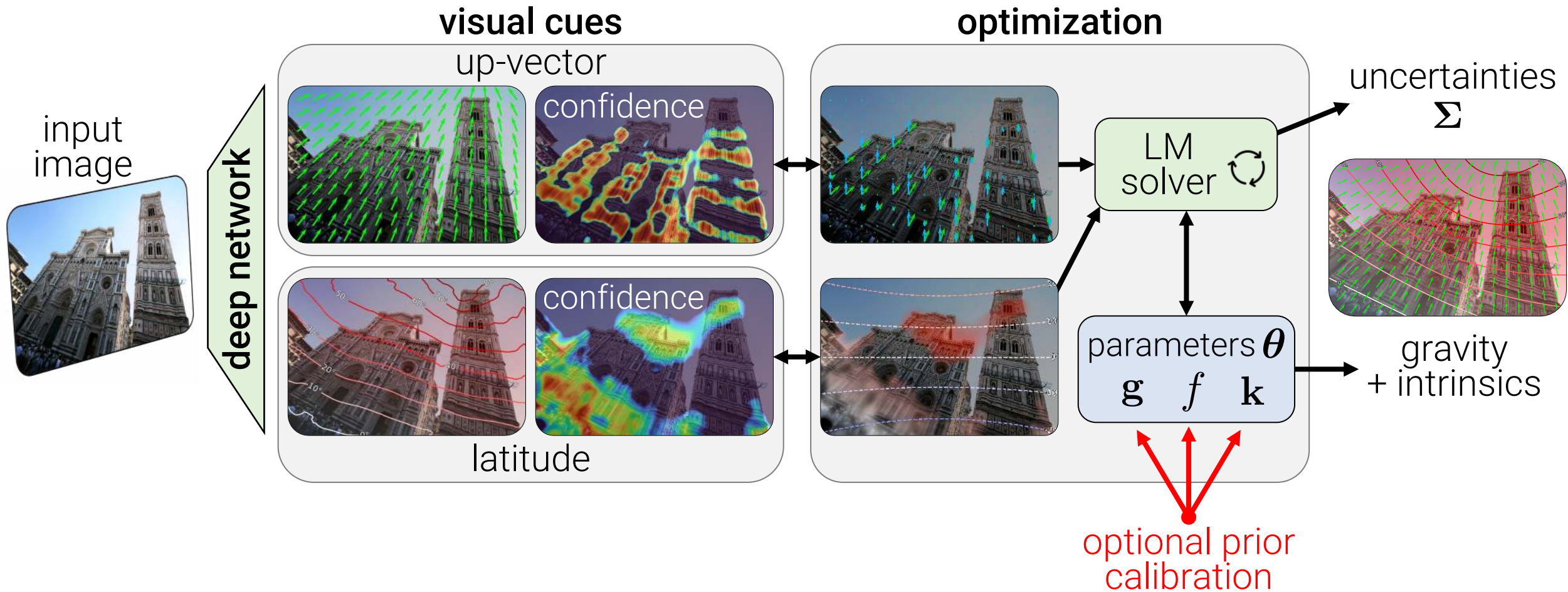
input

undistorted



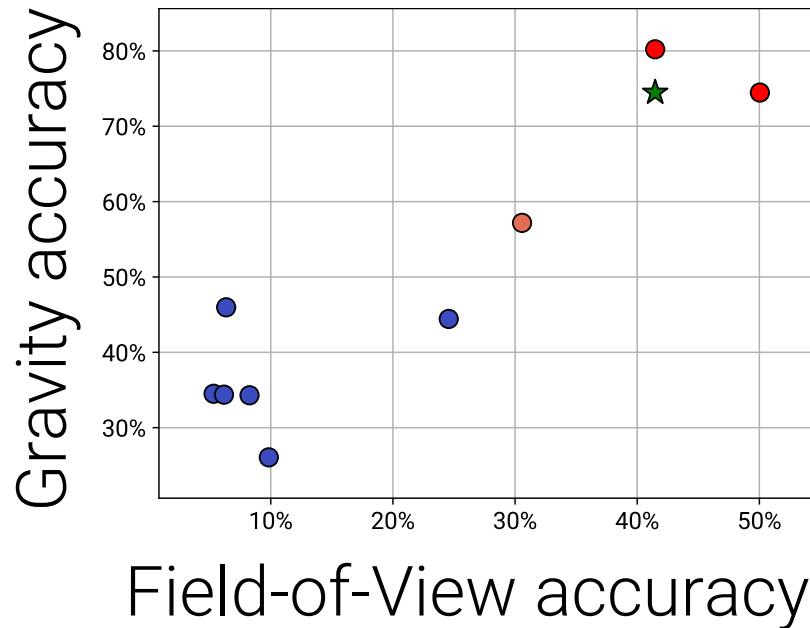
trained only on pinhole images

Practical benefits

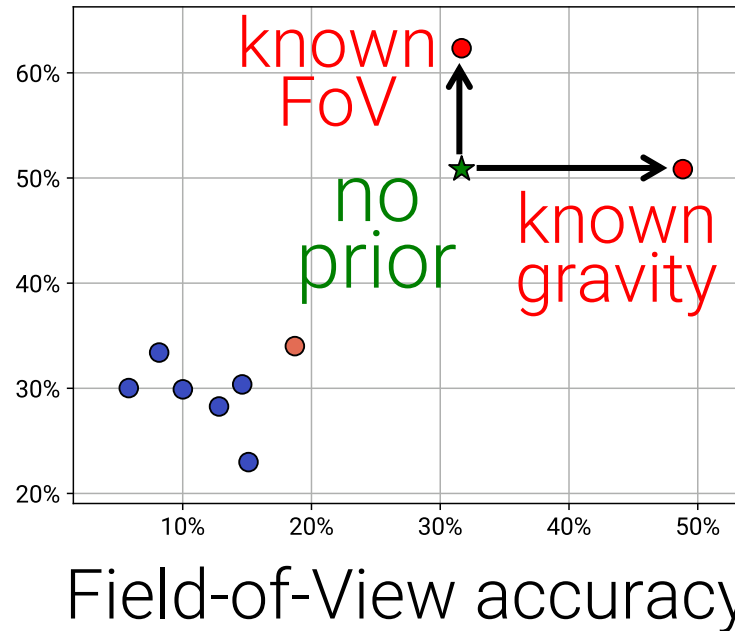


Leveraging partial priors

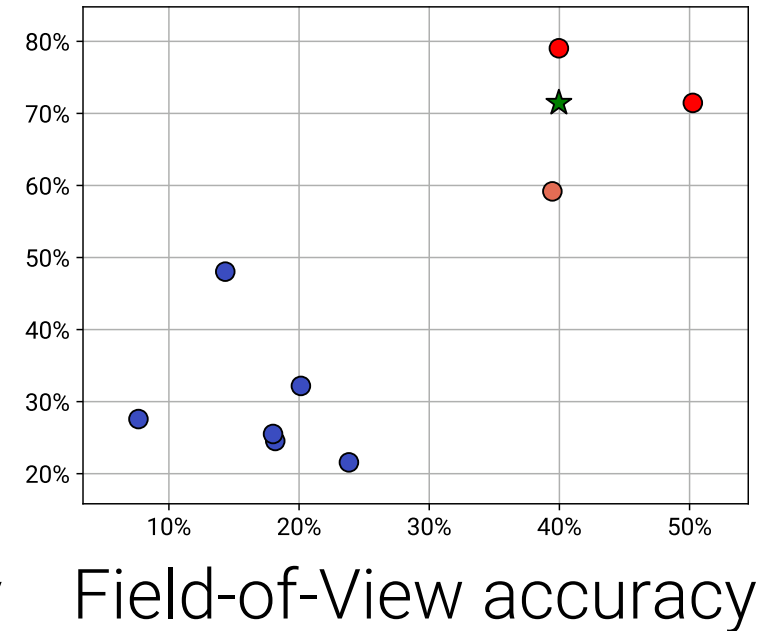
LaMAR - phones



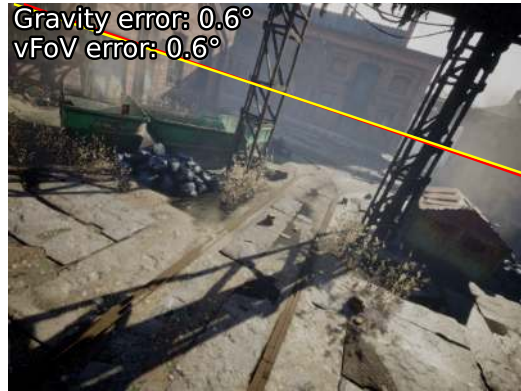
MegaDepth - outdoor



Stanford2D3D - indoor

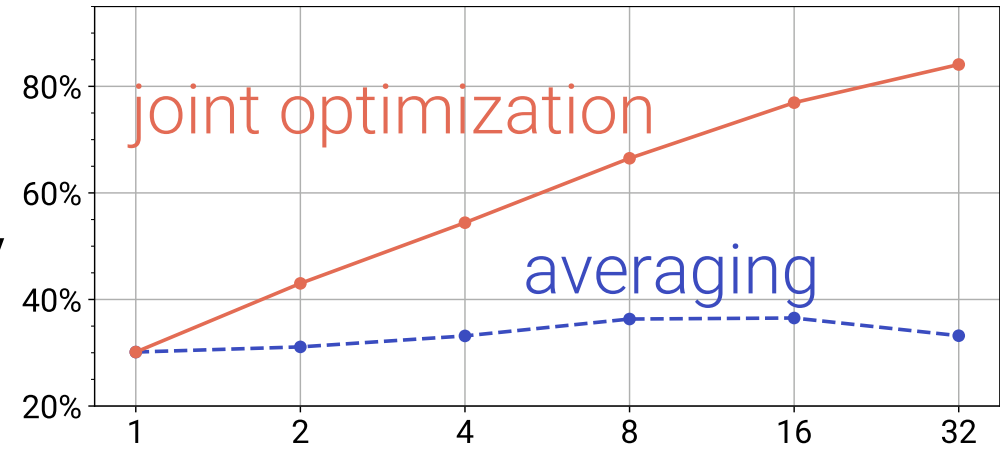


Multi-image optimization

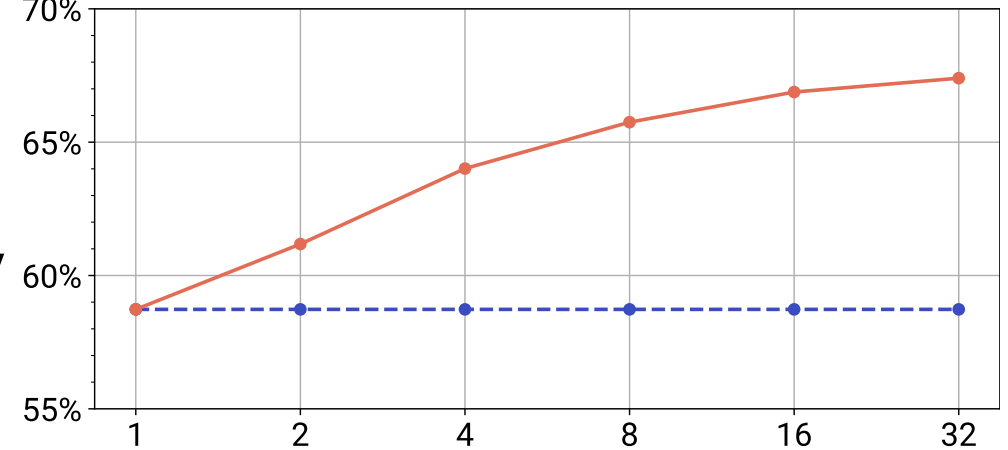


same camera
but different gravity directions

FoV
accuracy



Gravity
accuracy



number of images

GeoCalib = Learning + Geometry

More **accurate & robust** single-view calibration

Multiple benefits of **geometric optimization**:

- Uncertainties = interpretability
- Flexible camera models
- Optional partial priors
- Multi-image optimization



github.com/cvg/GeoCalib