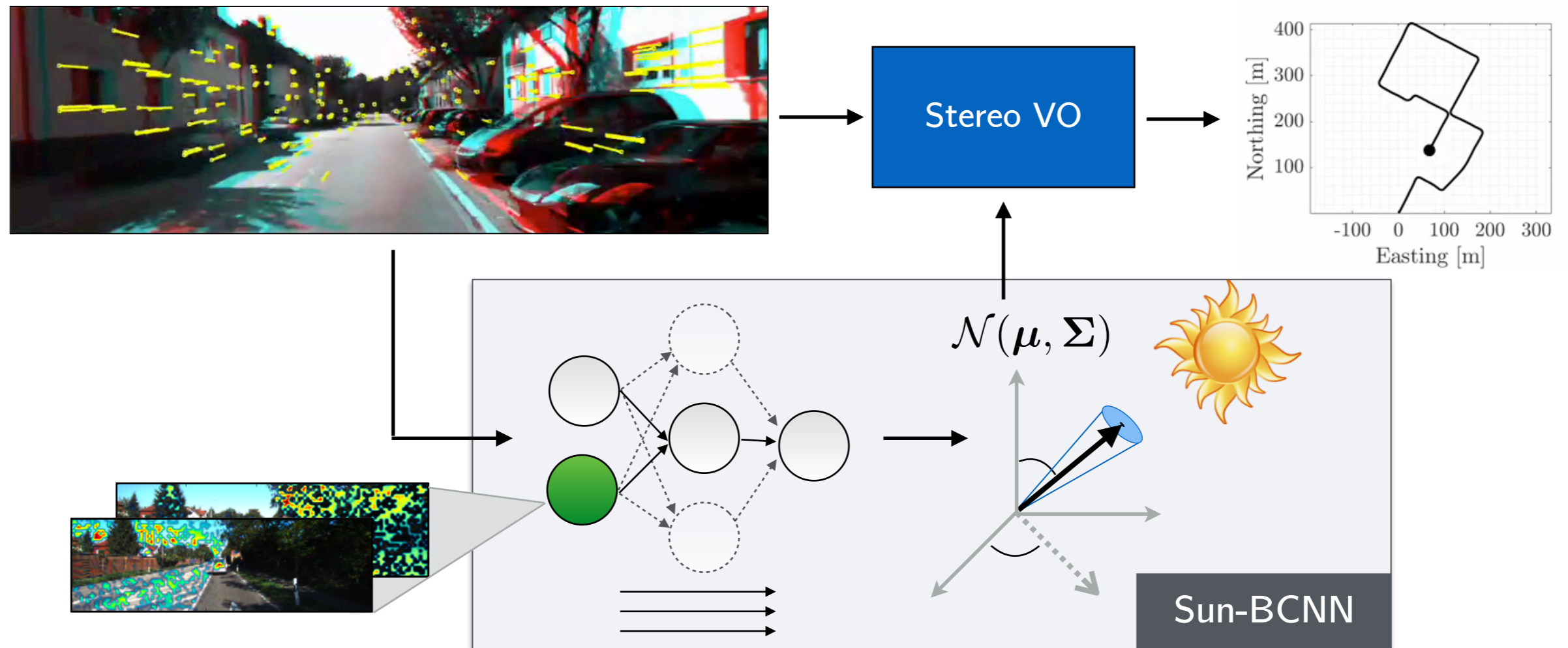


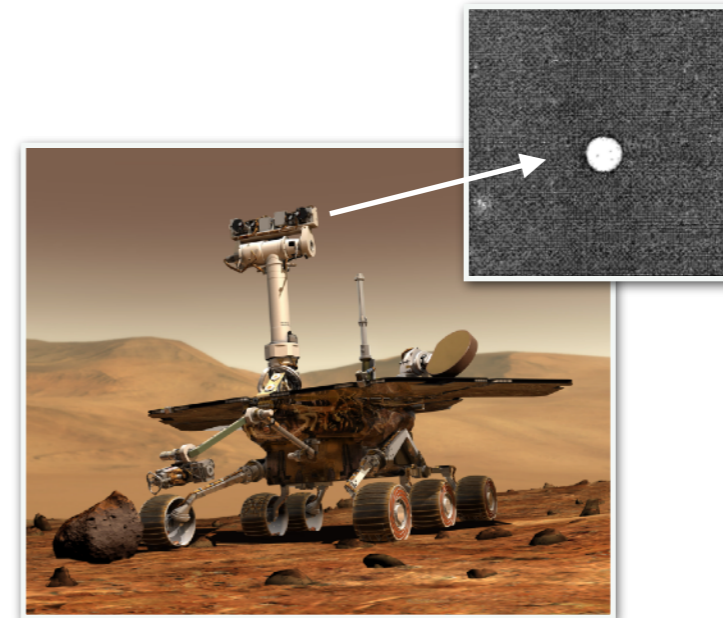
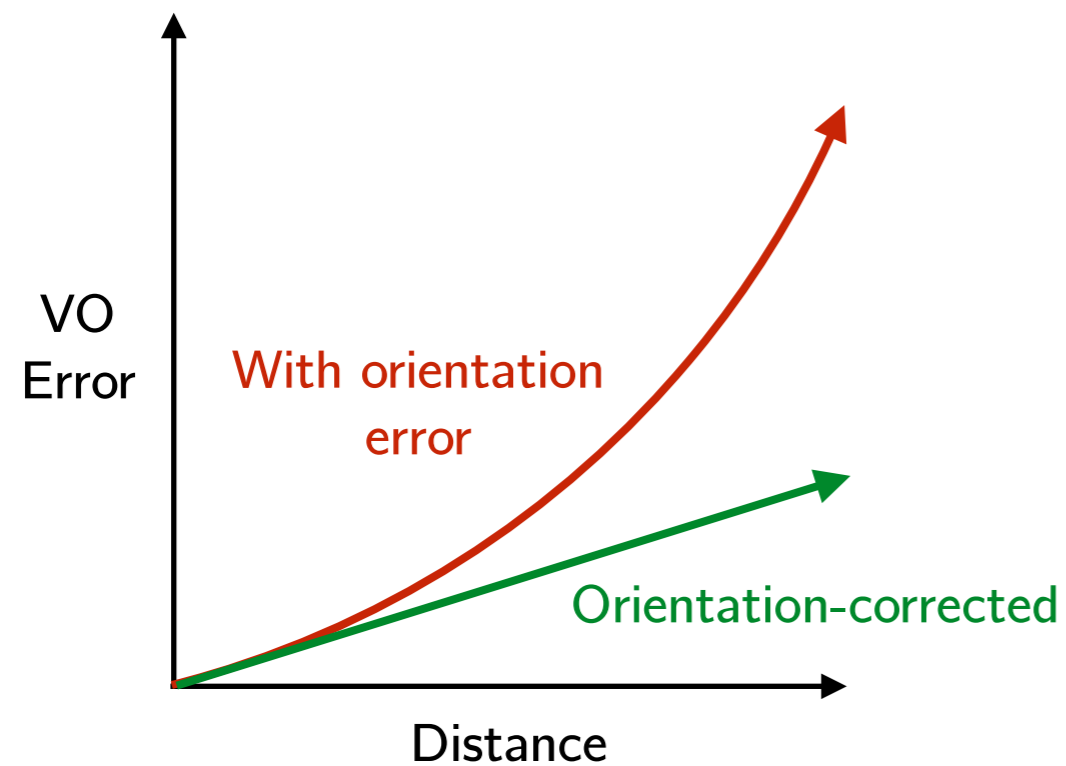
Reducing drift in VO by inferring sun direction using a Bayesian CNN

Valentin Peretroukhin, Lee Clement, and Jonathan Kelly

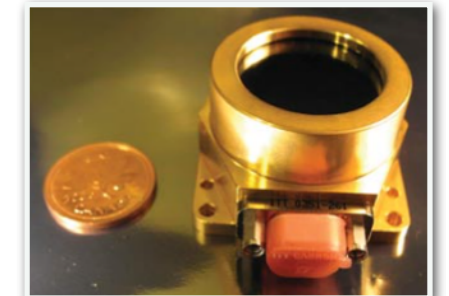


Sun-aided visual odometry

VO is a dead-reckoning technique and suffers from **super-linear error growth**, largely due to **accumulated orientation error**



Specially oriented camera (e.g., MERs)



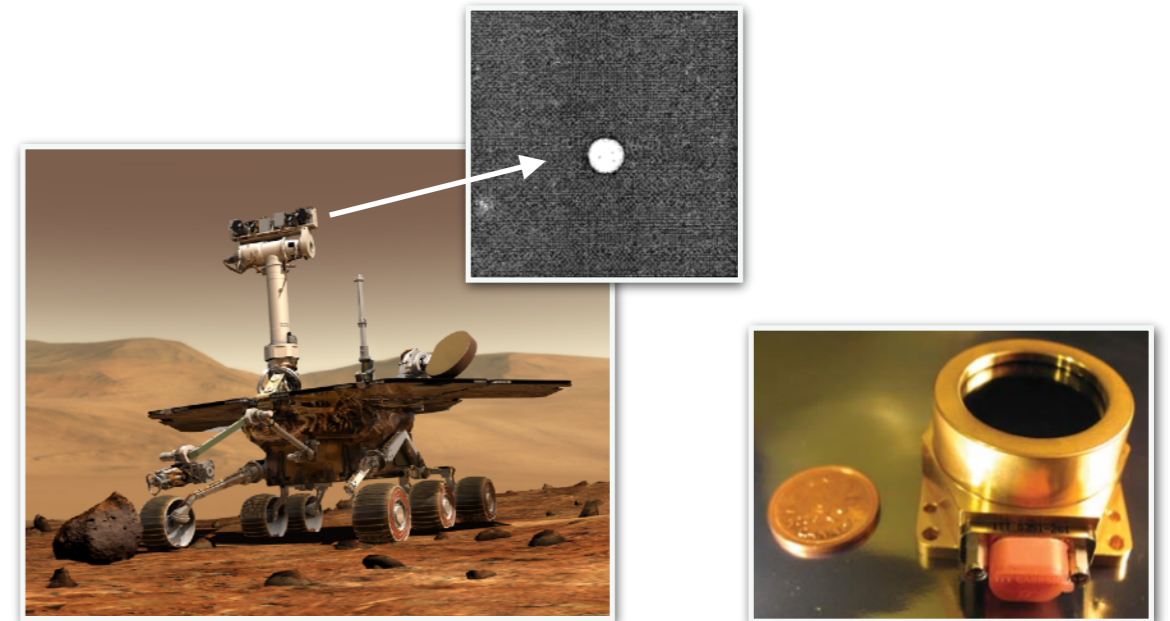
Specialized sun sensor

Drift can be reduced using **absolute orientation information** (e.g., observing the sun)

Simultaneous localization and... sun sensing?

Do we really need a hardware sun sensor or specially oriented camera?

In other words, do we need to look at the sun to see the sun?

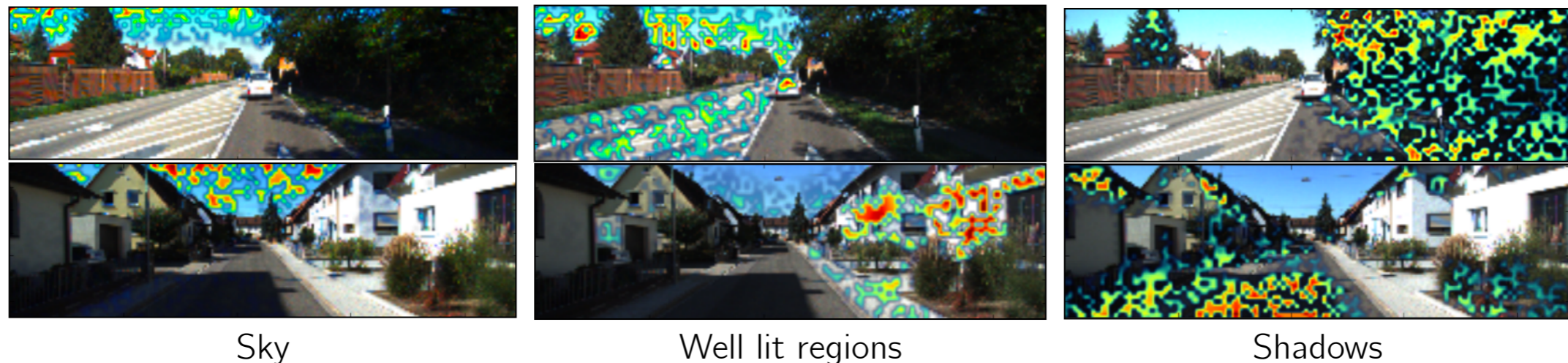
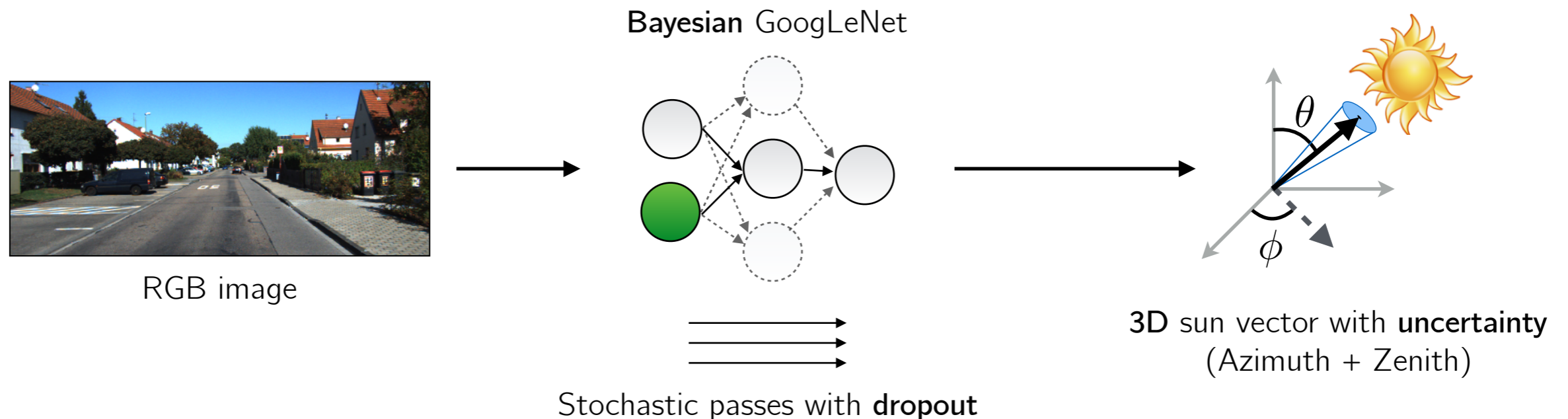


What if we could infer the direction of the sun from environmental cues?



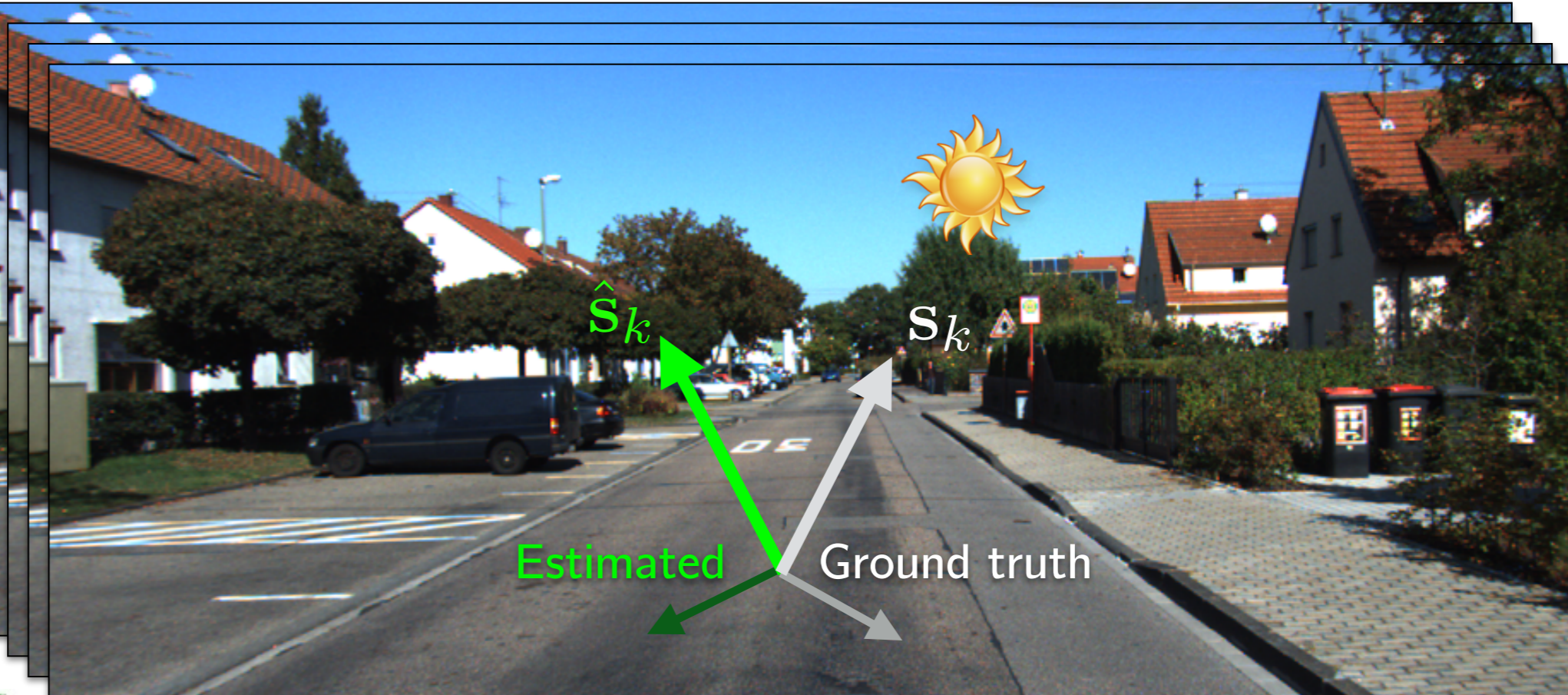
Sun-BCNN: A Bayesian CNN for finding the sun

We would like to estimate a **3D sun vector** and an associated measure of **uncertainty** by learning a model from data



Training: KITTI odometry benchmark

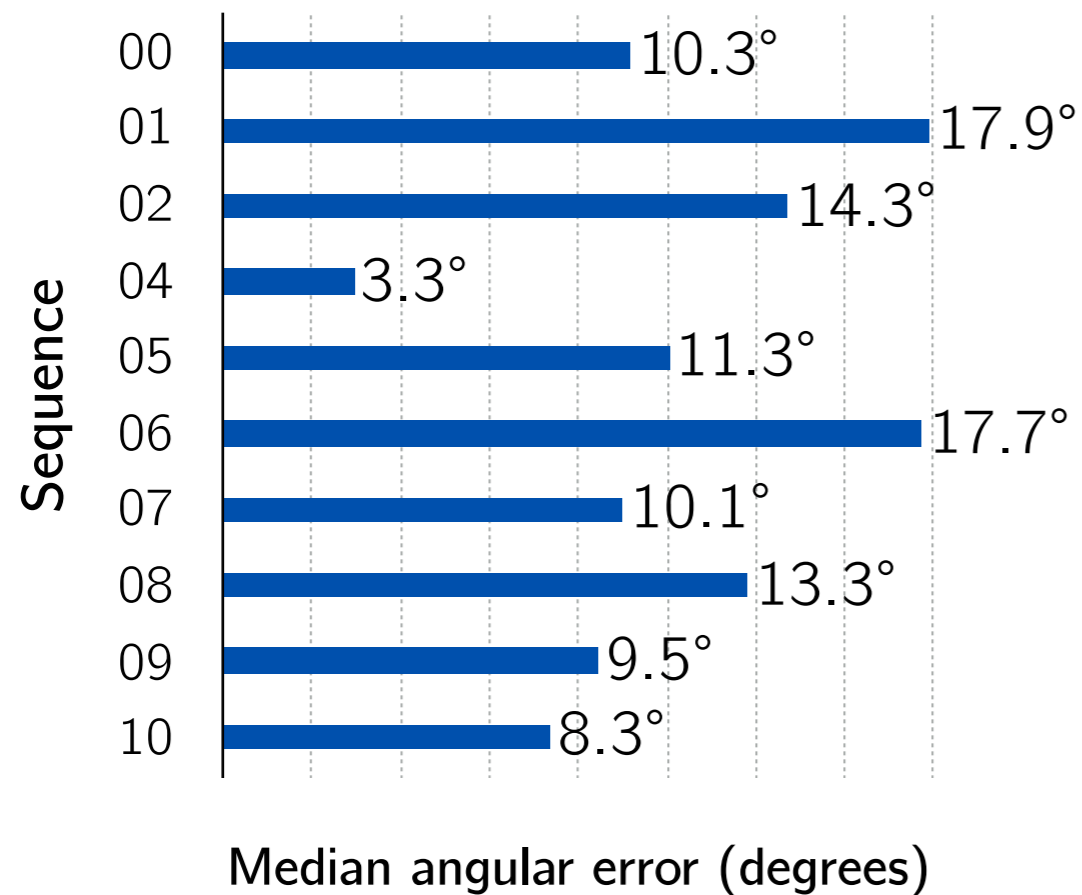
- 10 sequences
- 9/1 test/train split for each sequence
- 20k images per training set
- 1k epochs per training set



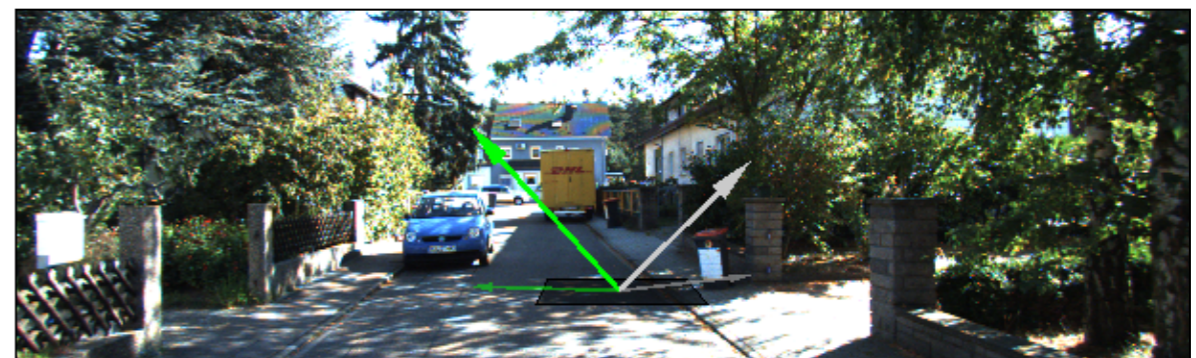
Ground truth? GPS-INS data from the KITTI dataset and a solar ephemeris model

Loss function? Cosine distance $\mathcal{L}(\hat{\mathbf{s}}_k) = 1 - (\hat{\mathbf{s}}_k \cdot \mathbf{s}_k)$

Testing: KITTI odometry benchmark



Sun-BCNN performs better with strong directional illumination cues

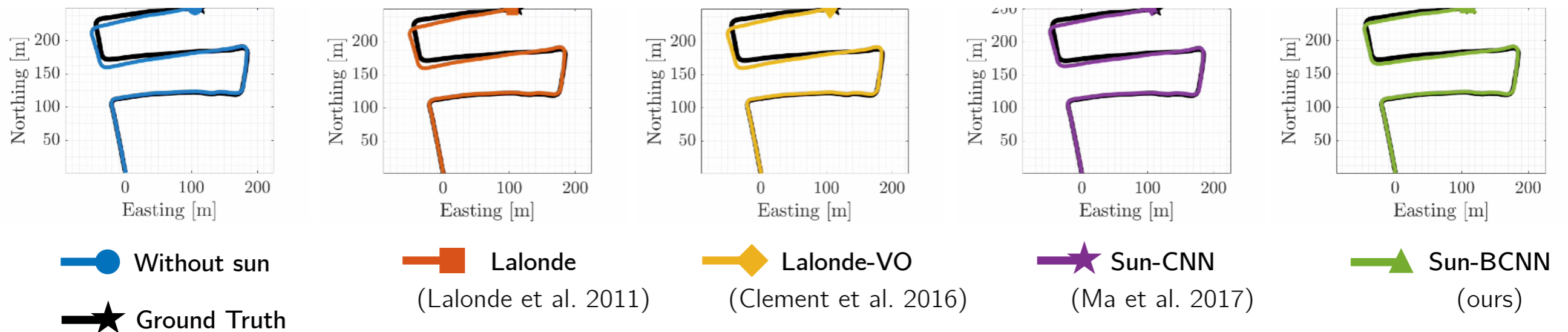


We consistently obtain **< 18 degrees median angular error**, but some sequences are better than others.

Visual odometry: KITTI odometry benchmark



VO Trajectories



Thank you!

Caffe implementation of Sun-BCNN
github.com/utiasSTARS/sun-bcnn



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