## Augmenting Computer Vision with expert knowledge and world-priors dr. Nicola Strisciuglio, Assistant Professor, Computer Vision and Machine Learning

Vision and motivation: Currently, computer vision models tend to have a larger and lager parameter space. Training them requires (hundreds of) millions of images (often paired with text captions), and large, expensive and energy-hungry computation facilities. Large models are powerful as they are able to disentangle effective representations from a large number of images. However, they learn relying mainly on assumptions on data quality. Our work is focused on including explicit expert knowledge or formalize priors from real-world tasks in the design and in the training process of convolutional architecture to facilitate the training of high-performance smaller models. We exploit neuroscientific findings and computational models, geometry-aware priors, and inspect the frequency characteristics of the data to exploit inductive bias and optimize training and application of CNN models.

Push-pull laver



disentangling motion and

generation of motion vectors.

representations and ensure motion-consistency

in the generated videos via an RNN-based

appearance

We aim at achieving models that can maintain high performance when deployed in scenarios with varying conditions.

## **UNIVERSITY OF TWENTE.**

Contacts

Thanks to all my students and collaborators for the brilliant work!



World- and camera-geometry aware priors for place recognition models. Traditional visual place recognition models are trained considering image similarity as a binary characteristic: two images are either depicting the same place or not. In practice, however, two images are x% similar. We exploit this prior concept and design a training strategy that involves world- and camerageometry priors to estimate a graded similarity ground truth. A novel Generalized Contrastive Loss function allows to exploit the graded similarity information during the training stage, which only takes few hours, in contrast with existing approaches that need about 45 days on the same hardware. We achieve new SOTA results on the Mapillary Street Level Sequences data set.





