Lecture: Deep Learning and Differential Programming

# **3.1 Computer Vision**

https://liris.cnrs.fr/christian.wolf/teaching



#### Goal: use <u>additional unlabeled</u> data for learning. Example:

#### Dataset 1:

Industrial application: Crop growth stage 3000 images + labels 16 classes, annotated

#### Dataset 2:

Industrial application: Crop growth stage 100 000 images <u>No</u> labels, <u>no</u> annotation



For each image: Label (e.g. which class among 16)

No manuel labeling Easy / cheap data acquisition!

Humans can predict the future.

If we learn how to predict the future ... can this help us to train a model which better understands the data?

<u>Self</u>-supervised learning is a vaiant of <u>semi</u>-supervised learning.



[T. Han, W. Xie, A. Zisserman, Video Representation Learning by Dense Predictive Coding, ICCV 2019]





#### Self-supervised learning Dense predictions



[T. Han, W. Xie, A. Zisserman, Video Representation Learning by Dense Predictive Coding, ICCV 2019]



Colors are labels (unused here during training!)

[T. Han, W. Xie, A. Zisserman, Video Representation Learning by Dense Predictive Coding, ICCV 2019]

## How about images?



[S. Gidaris, P. Singh, N. Komodakis, Unsupervised Representation Learning by Predicting Image Rotations, CVPR 2019]

The industrial perspective, by Varun Nair:

https://medium.com/@nairvarun18/from-research-to-production-with-deep-semi-supervised-learning-7caaedc39093

Can boost your performance. Probably can't solve your problem if performance with labeled data alone is very low.