

# Thesis Opportunities AA19/20

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# Reserach topics for a thesis

## Image processing and deep learning

- Anomaly detection
- Image Restoration and Enhancement
- Deep learning for unconventional data
- X-ray Multispectral Image Analysis

## Computer Vision

- Object detection by template matching
- Robust Model Fitting

## Change/Anomaly Detection in Datastreams

- Anomaly detection based on ensemble of histograms
- Change detection in datastreams

## Machine learning for health

- ML for Seismocardiogram and ECG signal analysis in wearable devices
- ML for sleep quality assessment
- Machine Learning for diagnosing nasal pathologies through CFD
- Biomarkers for Huntington Disease

In gray topics currently assigned

# Reserach topics for a thesis

## Stage Opportunities

- @STMicroelectronics (Agrate): 3 calls
- @Cleafy (Milano)
- @Fondazione Don Gnocchi (Milano)
- @Cisco (Vimercate)

In gray topics currently assigned

# These slides...

...provide both:

- A short overview of the research background (**Background** slides)
- Possible research directions along which to develop a thesis under our supervision (**Research directions** slides)
- Research institutions / companies involved in these research activities
- Stage opportunities



TAMPERE  
UNIVERSITY OF  
TECHNOLOGY



# The Team

You'll be supervised by myself and another experienced colleague.

Here we are:



*Giacomo Boracchi*



*Diego Carrera*



*Luca Frittoli*



*Filippo Leveni*



*Luca Magri*



*Simone Melzi*



*Andrea Schillaci*



*Diego Stucchi*

# For any enquiry

Just drop me an email and we can arrange a short meeting to discuss...

[giacomo.boracchi@polimi.it](mailto:giacomo.boracchi@polimi.it)

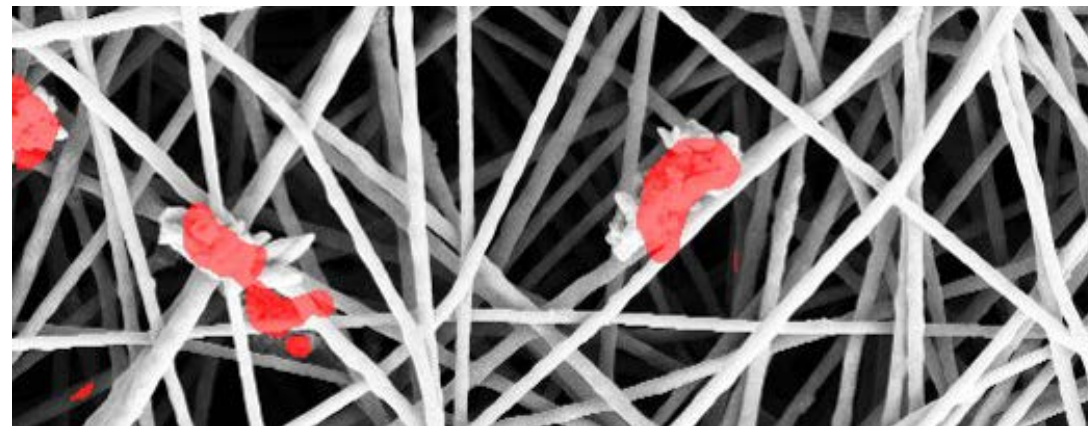
# Image processing and deep learning

## Background

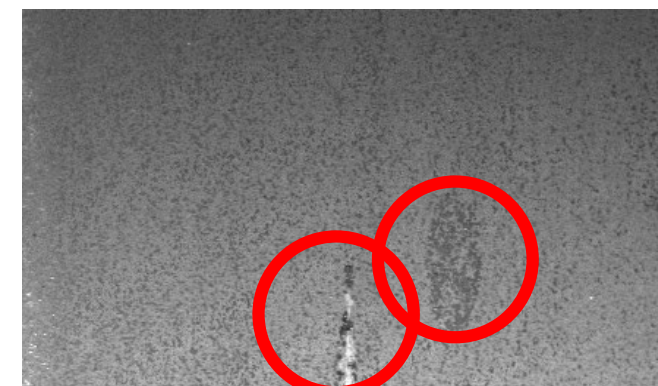
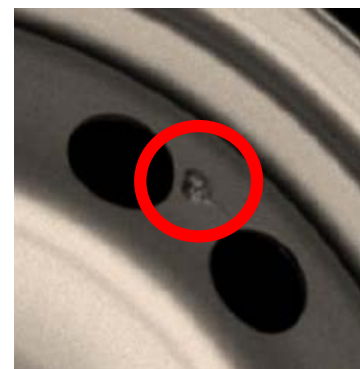
# Deep Learning for Anomaly Detection

Anomaly detection is a very general problem with applications ranging from: quality inspection to health and industrial monitoring

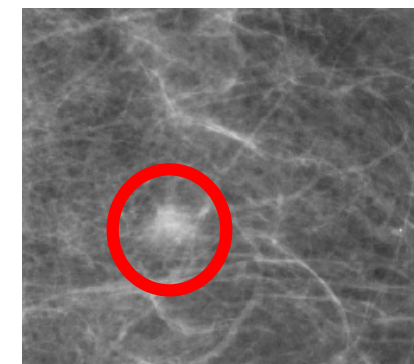
It is typically treated as an unsupervised problem, being anomalies unknown



*nanofiber production*



*manufacturing*



*mammograms*

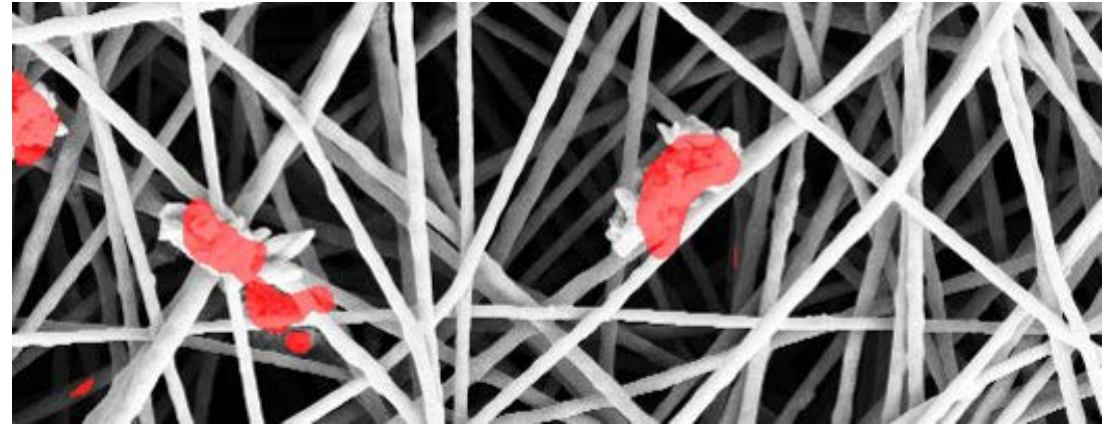


## Background

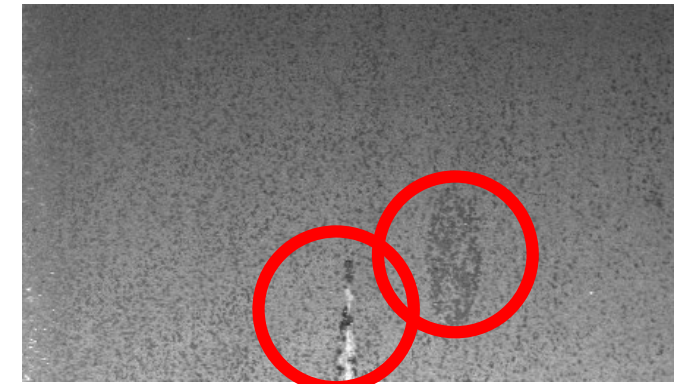
# Deep Learning for Anomaly Detection

Deep Learning (DL) represents the state-of-the-art in supervised tasks such as classification and semantic segmentation

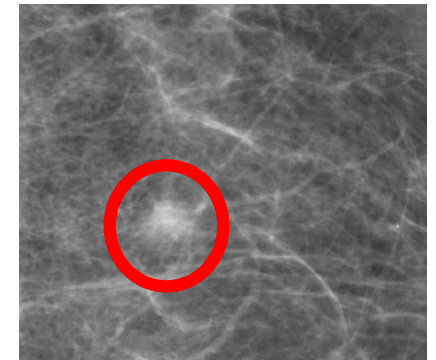
Unsupervised DL methods have been studied less, and the use of supervised techniques in industrial or health-related monitoring problems might not be straightforward



*nanofiber production*



*manufacturing*



*mammograms*

# GANs and Anomaly Detection

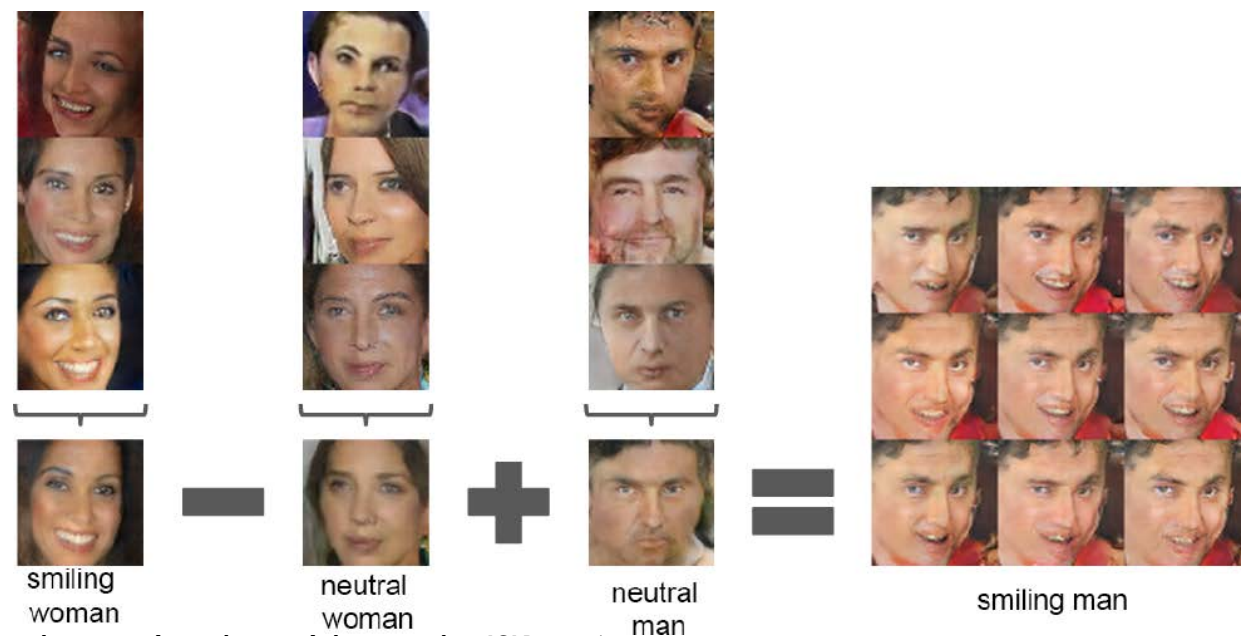
Generative Adversarial Networks (GANs): very effective generative models for images. These have been applied in a plethora of applications.

GANs are appealing for anomaly detection as they successfully parametrize the manifold where images live

Images generated after 5 training epochs



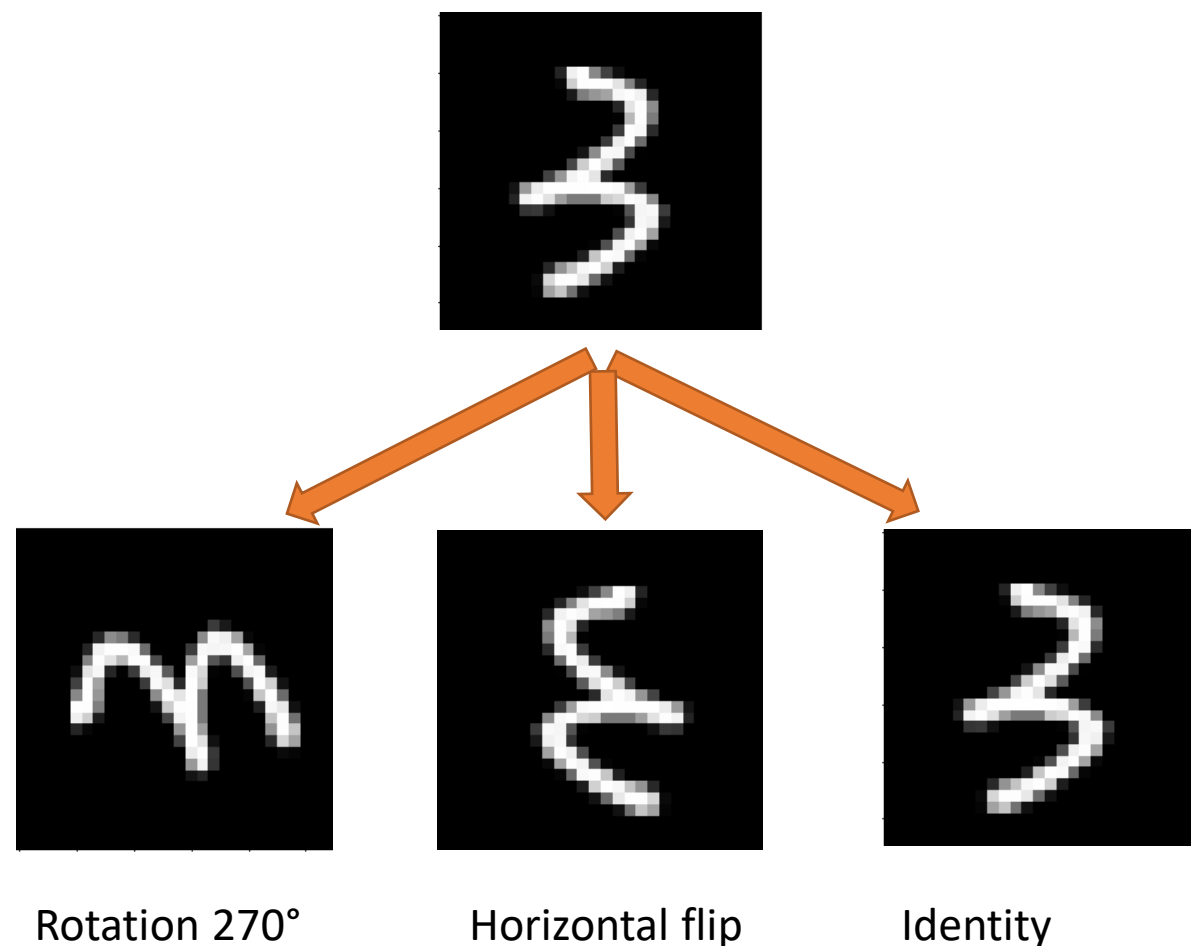
Image arithmetic: examples similar to word embedding



## Self-supervised anomaly detection

Geometric transformations can be used to create self-supervised datasets, labelled with the transformations being applied to the data.

Detecting anomalies based on the principle: *a classifier trained on normal images cannot classify geometric transformations applied to anomalous images*



## Thesis: Deep Learning for Anomaly Detection

### Research goals for a few thesis:

- Self-supervised methods for Anomaly Detection: automatically defining the best set of transformations to be applied
- Design **new training methods** for (deep) autoencoders that are specifically designed for AD.
- Investigate how **GANs can be used** to learn data representations and to augment training data, possibly in **sliding-window solutions**
- Develop AD methods able to deal with **inaccurate reference templates**

### Materials and Methods:

- Access to a server mounting **GPUs will be provided**
- **Annotated training sets** from industrial production
- Reference algorithm based on sparse-representations

# Image restoration and enhancement

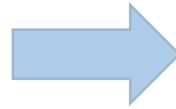
Noise always affects digital images and video frames

$$z(x, t) = y(x, t) + \eta(x, t), \quad x \in X, \eta \sim \mathcal{N}(0, \sigma^2)$$

Denoising algorithms provide  $\hat{y}$ , an estimate of the noise-free image  $y$ .



$z$



$\hat{y}$



## Image restoration and enhancement

Restoration algorithms have been widely investigated for their many applications in science and engineering

These often leverage **data-driven models** (including deep Convolutional Neural Networks, **CNN**) to provide effective representations of noise-free images.



# Image restoration and enhancement

Artifacts to be removed by restoration algorithms include noise and blur.

Both de-noising/de-blurring algorithms and the image formation process for this kind of artifacts have been widely investigated in the literature.

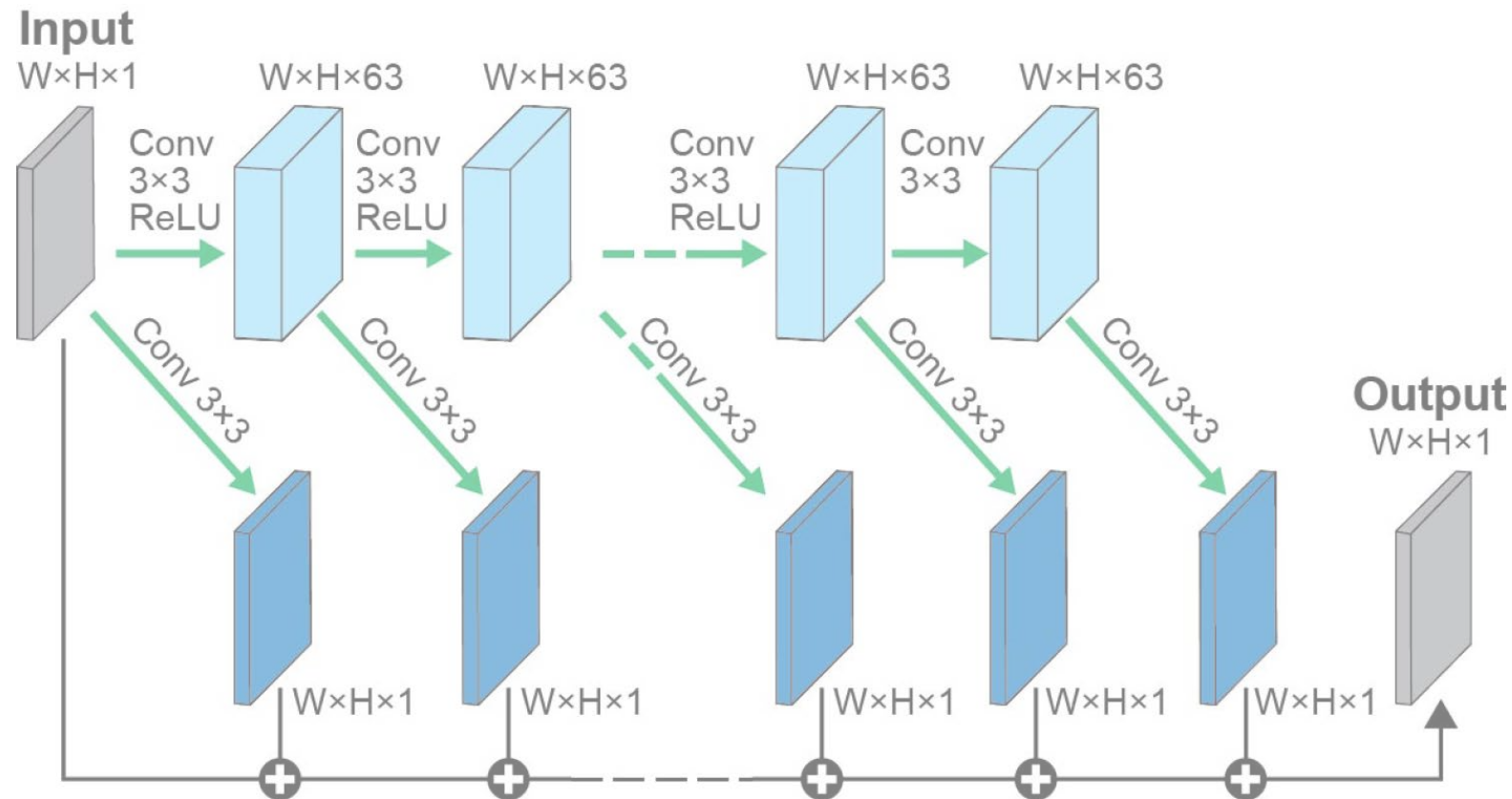
## Background



# Image restoration and enhancement

By leveraging image formation model and clean natural images it is possible to train powerful neural networks for

## Background





## Thesis: Image Restoration by data-driven (deep) representations

Goals (possibly more than one thesis)

- Study **data-driven representations (CNN, autoencoders)** of **noise-free images** and their use in combination with **key principles of denoising** algorithms such as:
  - Multiscale processing, self-similarity
- Consider denoising as a «mapping problem» and train a network to **learn transformations** from corrupted images to «clean» ones
- Force **rotation invariance** through specific layers such as in RotEqNet

**Materials and Methods:**

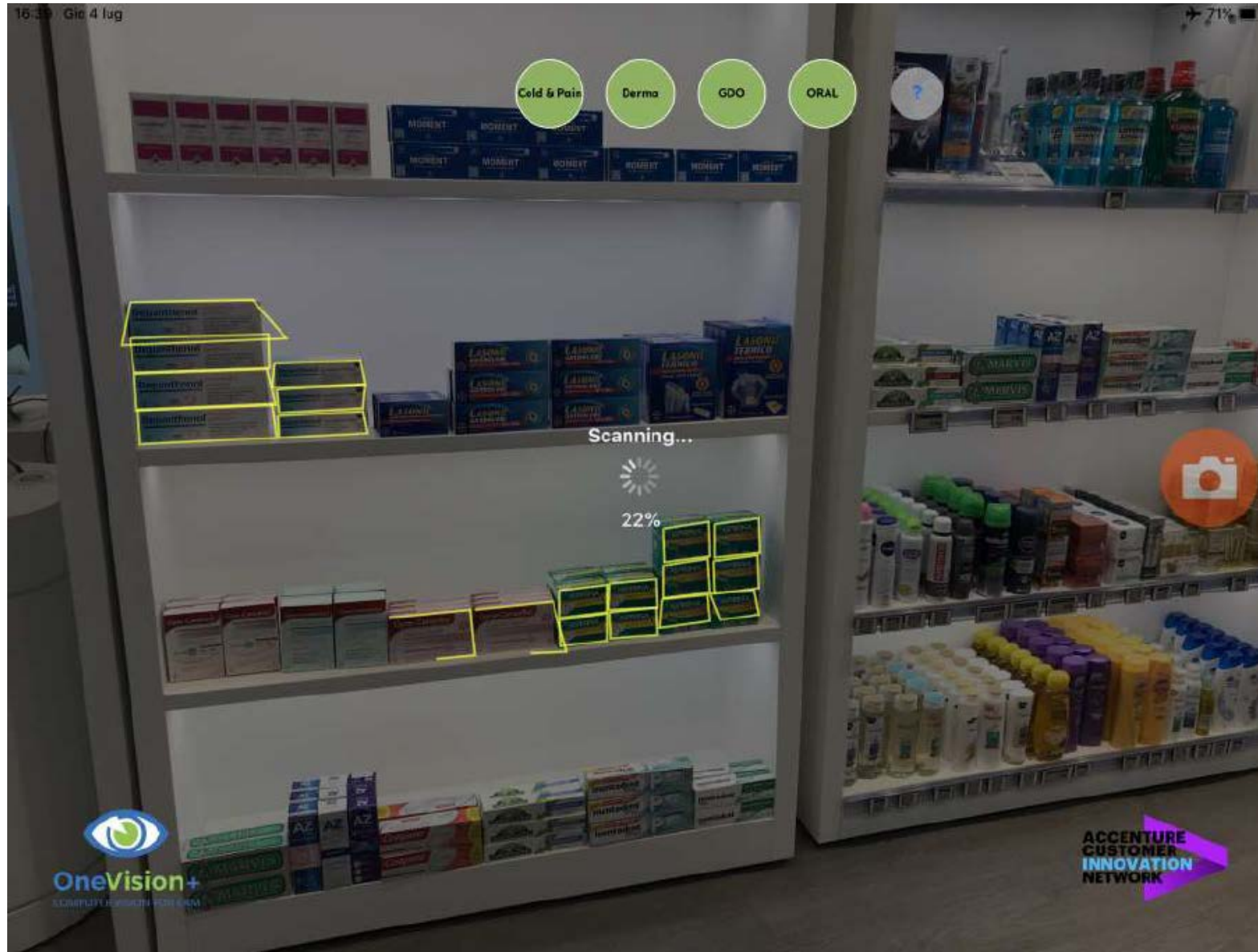
- Libraries for adding noise and corrupt images in realistic manner
- Access to a server mounting GPUs will be provided
- Reference algorithms to compare with the proposed solution

# Computer Vision

# Our template matching algorithms in action

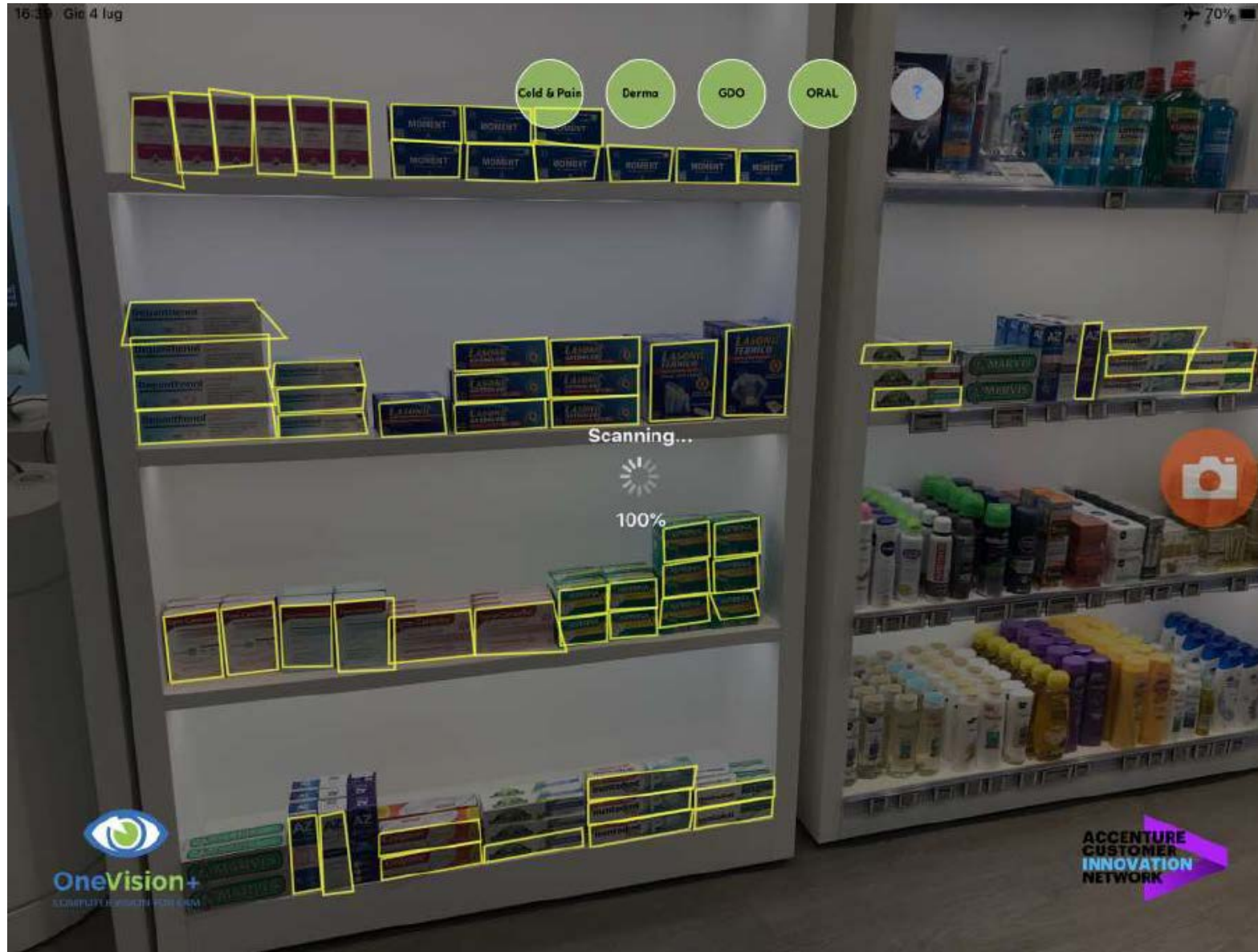


# Our template matching algorithms in action





# Our template matching algorithms in action



## Object Detection by Advanced Template Matching

Template matching is a very practical technique for finding all the occurrences of a template image in a given query image.

Deep-learning techniques are not very flexible, requiring re-training for new templates.

Traditional computer vision techniques (e.g. SIFT + RANSAC) are more practical but require an underlying model describing the transformation between the template and its occurrences in the query image.





## Thesis: Efficient template matching in large scale

Goals: Dynamically identify which template to search first when the number of templates is very large (e.g. a few hundreds):

- Studying state-of-the-art **template matching techniques** based on **image features** and **content based image retrieval**.
- Transfer learning of pre-trained models with very little samples in segmentation.
- Chromatic transformation to improve matching.
- Definition of **new templates** from a single image.



# Thesis: Efficient template matching in large scale (cnt)

**Goals:** Dynamically identify which template to search first when the number of templates is very large (e.g. a few hundreds):

- Implementation of learning-based criterion to identify the template that will be most likely found in a region of the query image.
  - Clustering of template images to speed up the retrieval
  - Learning a latent representation of templates

## Materials and Methods:

- Access to a server mounting GPUs.





## Thesis: Template matching for non-rigid distortions

Non-rigid distortions can not be modeled to enable RANSAC.

**Goals:** Design efficient clustering algorithms for feature matches that identify multiple, distorted, instances of the template in the query image:

- Studying state-of-the-art **template matching techniques** based on **image features**.
- Design of **learning-based** feature matching schemes.
- The design of an innovative technique to **group feature matches** that refer to the same template instance in the query image.

**Materials and Methods:**

- Access to a server mounting GPUs.
- Dataset from our industrial partner.

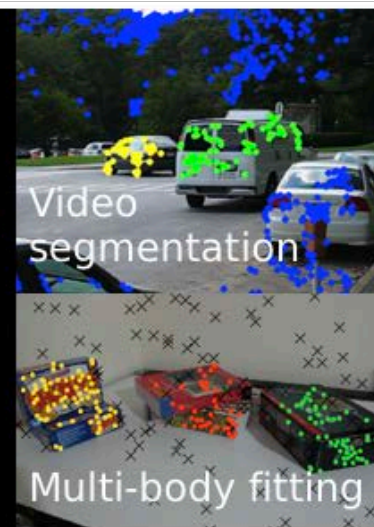
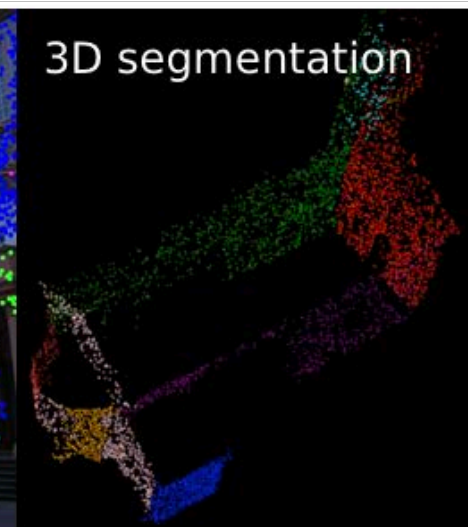
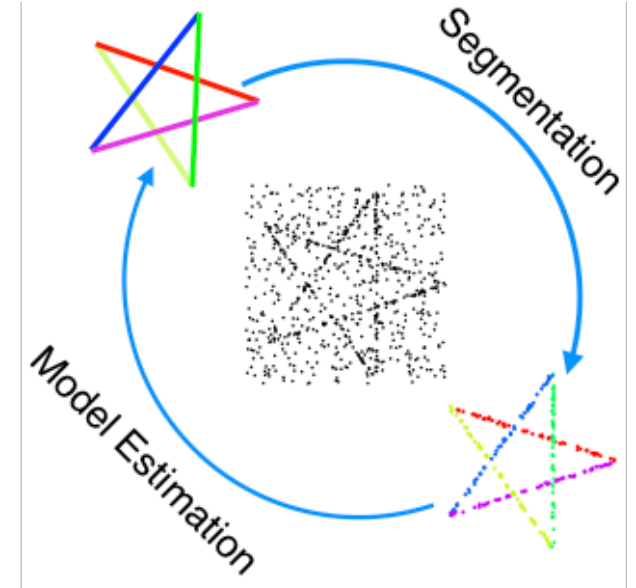


# Robust Model Fitting in Computer Vision

**Goals:** Fitting one or multiple instances of a geometric model – also called "structures" – to measured data, which is invariably contaminated by noise and outliers.

## Research Challenges:

- outliers
- pseudo-outliers
- chicken- $\hat{c}$ -egg-dilemma
- ill-posed problem
- scale estimation
- model selection



# Robust Model Fitting in Computer Vision

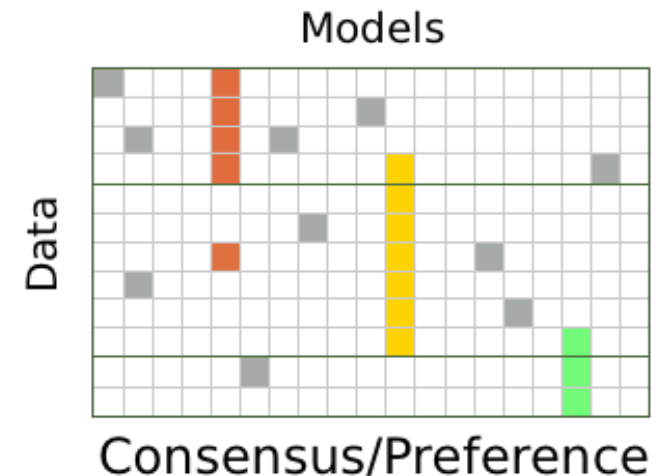
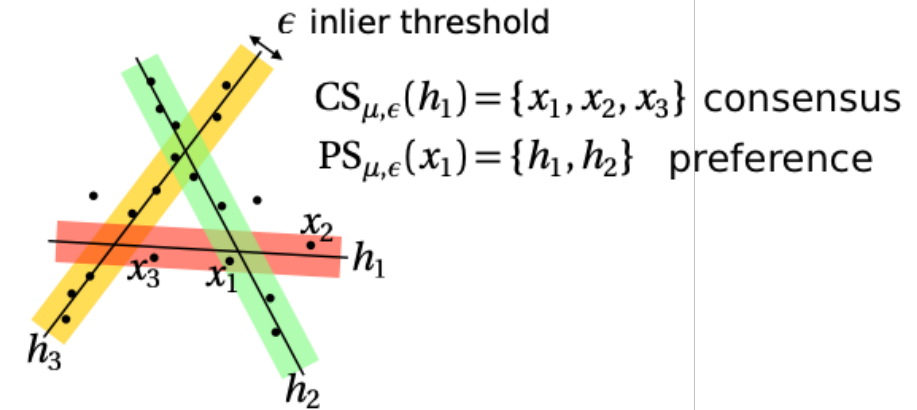
Possible approaches (among others)

**Consensus analysis:** (e.g. Ransac, Hough transform)

1. represent models (columns of the matrix)
2. count the number of points explained by each model (consensus)
3. maximize consensus

**Preference analysis:** (e.g. Residual Histogram Analysis, T-Linkage, Higher order clustering)

1. represent points (rows of the matrix)
2. each point “votes” for the preferred models
3. clustering of points preferences



# Thesis: Deep Multi-model fitting

## Research goals for exploratory thesis:

investigate how deep learning can be used to boost multi-model fitting algorithms:

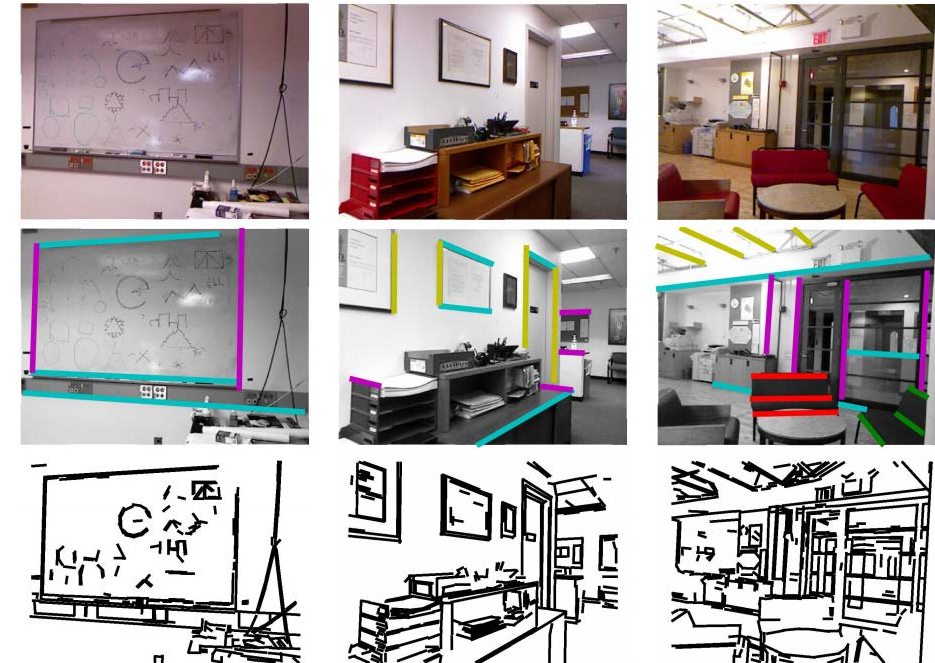
- [1] exploits unsupervised learning to improve RanSaC, it is possible to extend this framework to the case of multiple models?
- In [2] deep learning is used to guide the sampling towards promising models, it is possible to integrate this approach in classical multi-model fitting algorithms?

## Materials and Methods:

- Annotated datasets for two-view body segmentation and video segmentation.

## Some reference to start from:

1. Probst et Al. Unsupervised Learning of Consensus Maximization for 3D Vision Problems, CVPR 19
2. Kluger et Al. Consac: Robust Multi-Model fitting by Conditional Sample Consensus, Arxiv.





# Thesis: Dealing with outliers

## Research goals for a few thesis:

- Develop methods able to **accurately remove outliers** for geometric computer vision tasks
- Study and compare existing **outlier rejection techniques** (robust statistic, low rank decomposition, NFA, topological)
- Investigate how **the sparsity of data/model** representations can be exploit to detect outliers.
- In situations where there are many outlier points, any wrong model would achieve similar consensus. Correct models can be detected as models have anomalous consensus

## Materials and Methods:

- Annotated datasets for two-view body segmentation and video segmentation.

## Some reference to start with:

- Tepper and Sapiro, “Fast L1-NMF for Multiple Parametric Model Estimation”, 2016

## Thesis: Robust estimation in an uncertain world

### Research goals for a few thesis:

- Investigate how the inherent uncertainty in data location, can be integrated in the estimation of geometric structures to obtain a more efficient robust fitting algorithm.
- Study the use of randomized model verification to characterize the ‘non-randomness’ of a solution.
- Explore the extension of this result to the case of multiple models.

### Materials and Methods:

- Annotated datasets for two-view body segmentation and video segmentation.
- Develop a robust feature matching application

### Some references to start with:

- Tordoff and Cipolla, “Uncertain RanSaC”, Conference on Machine Vision Applications 2005.
- Raguram, Frahm, and Pollefeys. "Exploiting uncertainty in random sample consensus." ICCV 2009.

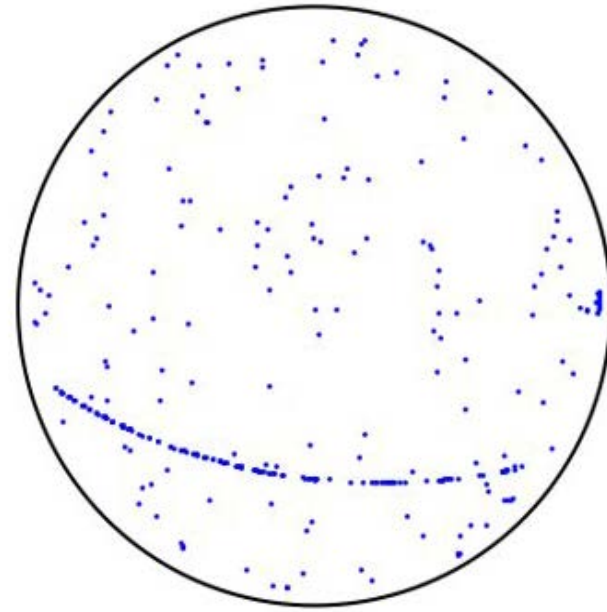
# Nonconventional imaging modalities

## Deep Learning on unstructured data:

Most DL algorithms are based on **discrete convolution**, an operation defined for data with an underlying **grid structure** (e.g. images)

Industrial data might lack such a structure. For instance, data might consist of a set of **3D points** or **measurements in scattered locations** (e.g. point clouds, defect maps, sensor networks acquisitions)

A popular solution is to **discretize** the data points and force them into a grid structure. However, this procedure might cause a **loss of information**



*A Wafer Defect Map (WDM) is a list containing the coordinates of the defects on a wafer. Defects are displayed on a huge grid (20,000 x 20,000) that would be impossible to handle as an image*

*Point clouds are measurements scattered in the 3D space from LIDAR or RGB-D cameras. Acquisitions from sensor networks can also be conveniently modeled like that*





## Thesis: Continuous Convolutions to learn from unstructured data

**Goals:** develop new DL methods to handle and learn directly from sets of points. In particular, this project will address:

- The study of **continuous convolutions**, where the kernel is a **continuous function** instead of a discrete matrix
- The design of a **continuous convolutional** layer
- The development of a DL architecture based on continuous convolutions and validation of the solution

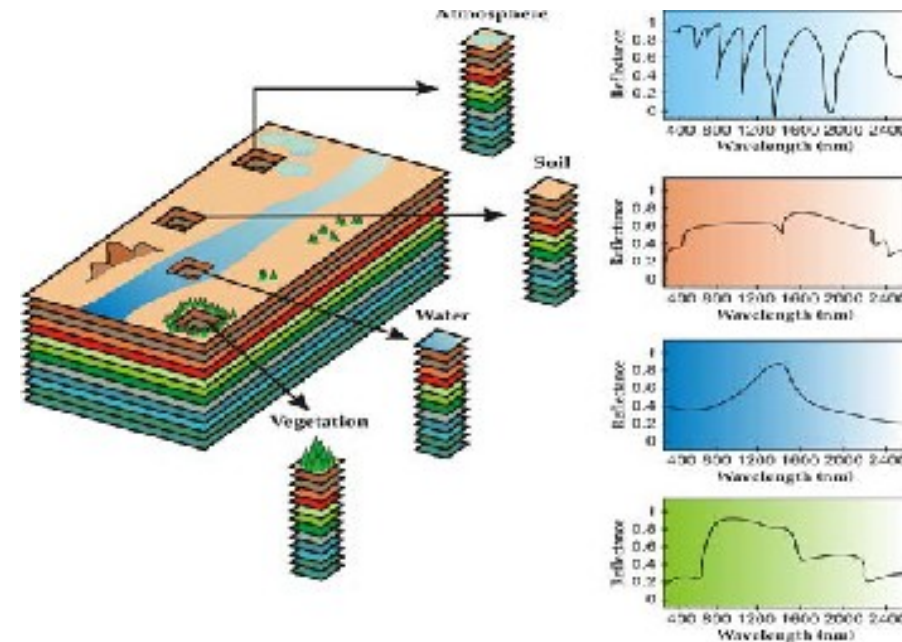
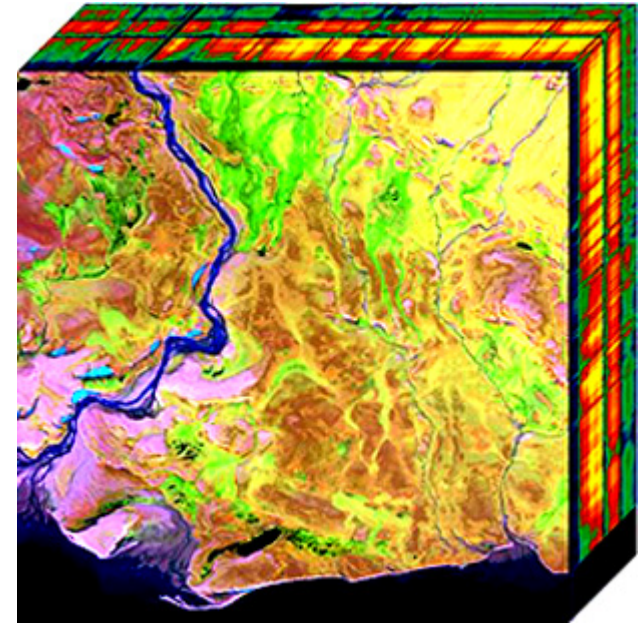
### Materials and methods:

- Access to a server mounting GPUs will be provided
- Annotated from sparse wafer defect maps, public datasets point clouds
- This thesis can be possibly addressed during an internship at STMicroelectronics

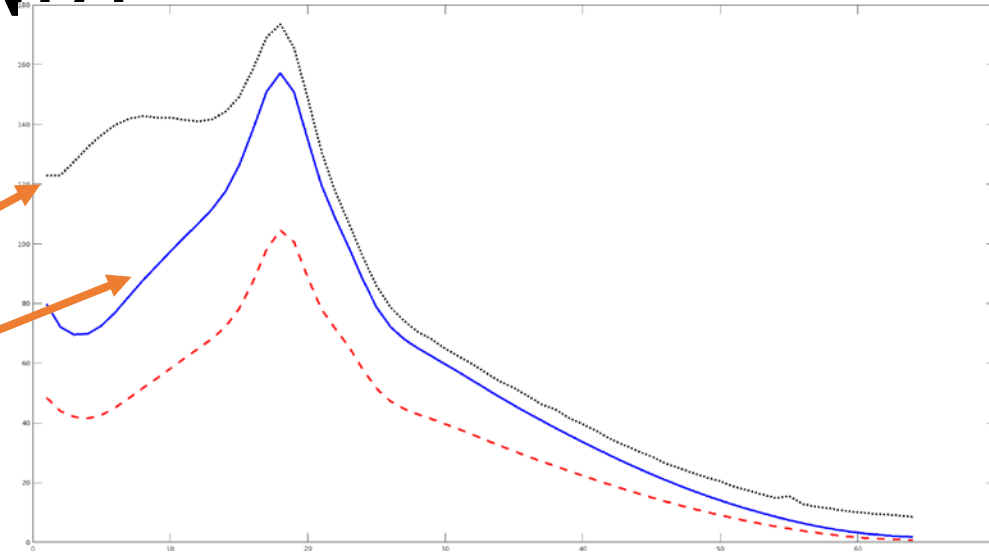
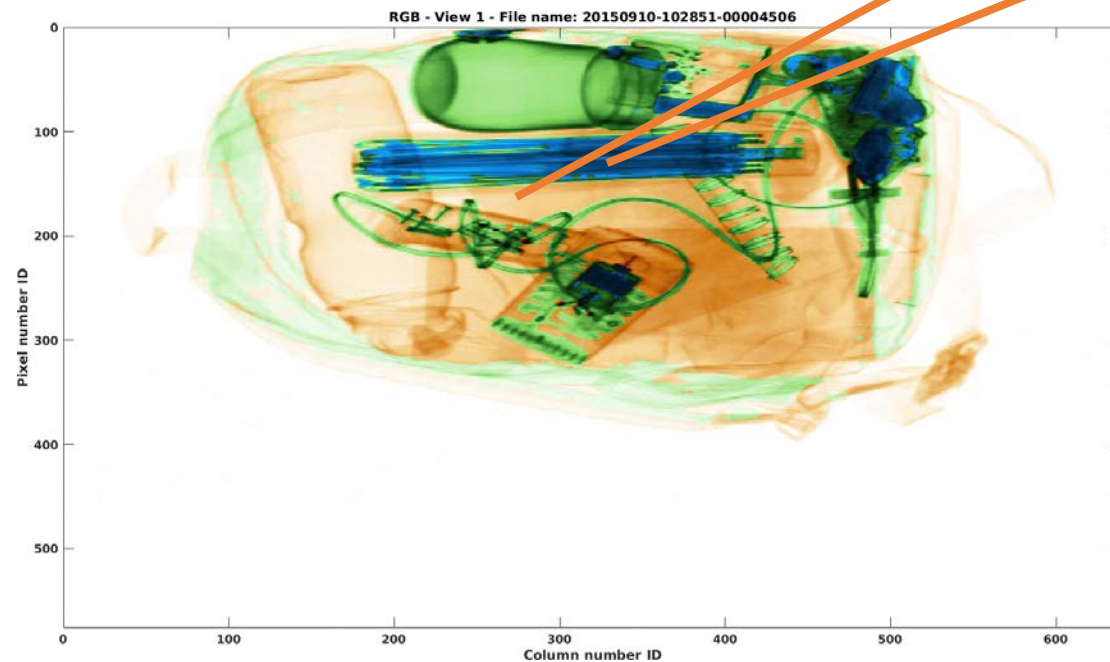
## Multispectral Image Analysis

Multispectral and hyperspectral images gather data from multiple spectral bands. Each pixel, is thus associated to a spectrum of a few tens / few hundreds of intensity values, each representing a certain wavelength.

Relevant challenges in this field are classification, recognition and unmixing i.e. separating an acquired spectrum as the sum of few basic spectra (the endmembers)



# Spectral Unmixing for X-ray NDT



Collaboration with



## Thesis: Multispectral Image Analysis

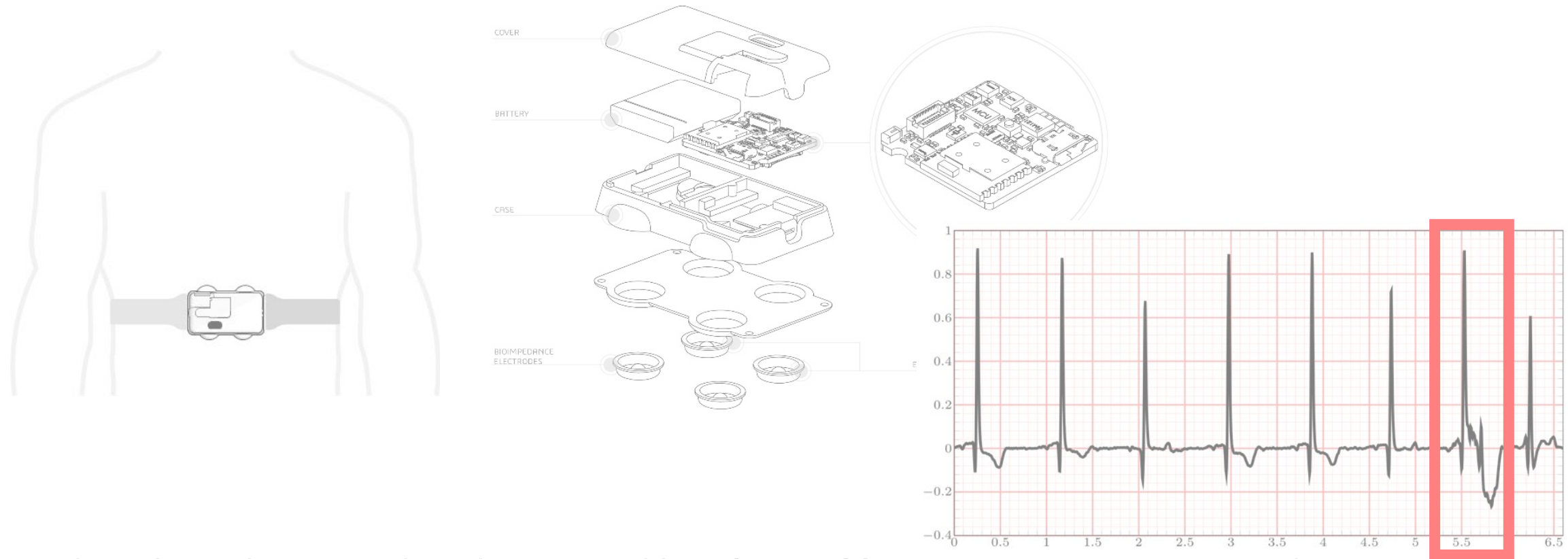
### Goals:

- Develop advanced classification and recognition algorithms for satellite imaging that combine additional information like GIS data and proximity constraints.
- Design spectral unmixing algorithms able to identify known materials in x-ray multispectral images acquired from baggage inspection systems.

**Materials and Methods:** multispectral images from opendigital and from X-ray machinery are available for testing unmixing and classification algorithms

# Machine Learning for Health

# Online and long-term ECG monitoring in wearables



We have been designing algorithms to enable **online and long term ECG monitoring** on low-power wearable devices. Our algorithms are based on **data-driven models** to automatically **detect anomalies** (e.g., arrhythmias), these models undergo a **domain adaptation** procedure to track heart rate variations. The model is based on a **learned and user-specific model** of heartbeats.

IJCAI Demo webpage: [http://home.deib.polimi.it/carrerad/IJCAI\\_2018\\_Demo.html](http://home.deib.polimi.it/carrerad/IJCAI_2018_Demo.html)

## Thesis proposal on ECG Monitoring

### Goals:

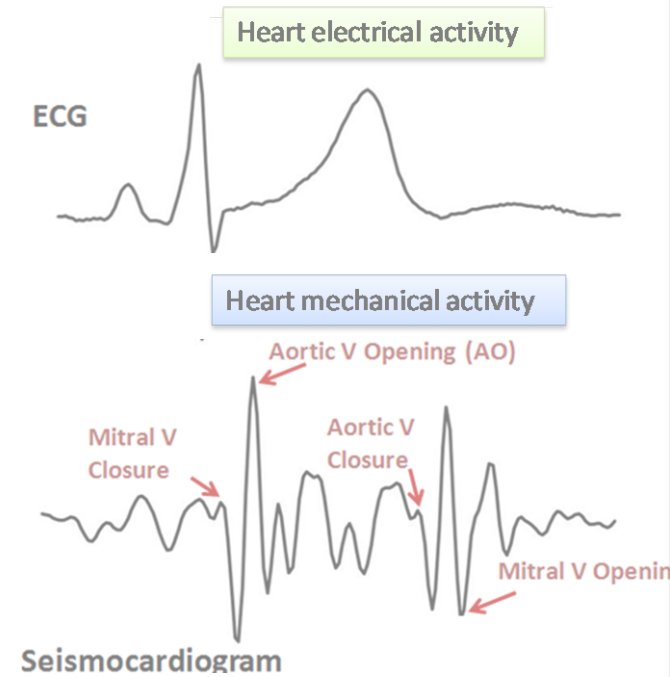
- Train 1d-CNN / deep RNN to classify heartbeats and identify arrhythmias
- Design efficient algorithms to enable online monitoring
- Design easy-to-train transfer learning algorithm that require few training data. In particular, adapt a pre-trained general-purpose models to:
  - analyze ECG signals from each specific user
  - Operate at different heart-rates
  - Operate in presence of motion artifacts

**Materials and Methods:** We currently have a **prototypal wearable device** equipped with ECG and Bioimpedence leads as well as MEMS accelerometers. Annotated datasets from multiple users have been prepared.



## Machine Learning for Seismocardiogram analysis

- Seismocardiogram (SCG), records micro-accelerations of the chest wall due to the heart movements, and it's a very informative signal since, w.r.t. traditional ECG:
  - SCG provides a direct measure of the **heart mechanical activity**, and not just the heart electrical triggers (assessed by the ECG).
  - The analysis of SCG is a quite new research, and efficient **automatic algorithms for segmenting, classifying and identifying anomalies** in this signal have still to be investigated.
- Recently, Fondazione Don Gnocchi developed a wearable system for SCG monitoring, which have been used to monitor patients in clinics, and also in the International Space Station to monitor astronauts during their sleep.
- See the video at <http://ow.ly/7z7Q30mg11K>





# Thesis proposal: ML methods for Seismocariogram analysis

## Goals:

- Develop [data-driven algorithms](#) for the automatic localization of patterns (associated with salient moments of the cardiac cycle, such as the opening and closure of valve) within the SCG waveform
- Develop patient-specific [ML models](#) to detect anomalies in the SCG signal.

## Materials and Methods:

- The activity will be carried out in collaboration with the [Wearable Sensor and Telemedicine Laboratory](#) coordinated by Ing. Marco Di Rienzo, of the IRCCS [Fondazione Don Carlo Gnocchi](#) Hospital in Milano.
- The [SCG recording device](#) (SeisMote) as well as annotated datasets from patients will be available for the thesis.

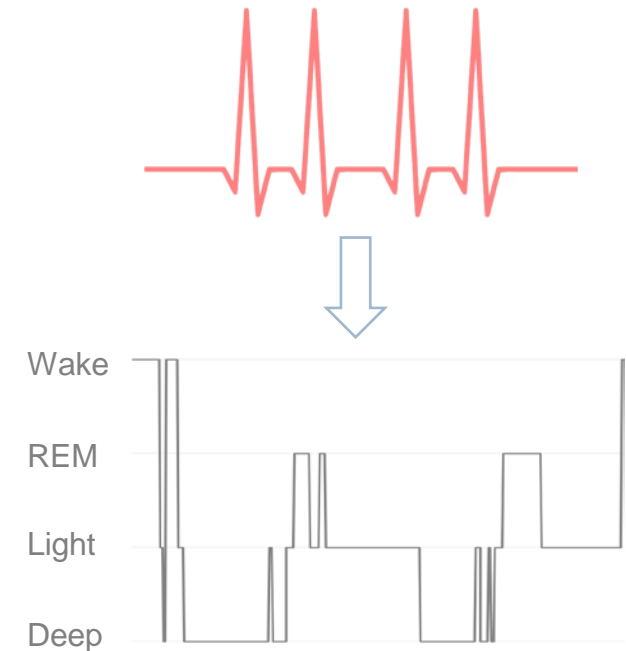
# Sleep quality assessment

**Goal:** defining an indicator of **sleep quality** starting from data related to users' vital signs gathered by means of a smart t-shirt.

**Research challenges:**

- Reconstruction of the hypnogram (sleep graph) by using vital signs
- Data quality
- Adaptation: sleep quality depends on users' profile and habits

**Materials and Methods:** We currently have a smart t-shirt equipped with sensors for sleep monitoring and public annotated datasets



*Research in collaboration with Prof. Cinzia Cappiello (DEIB)*

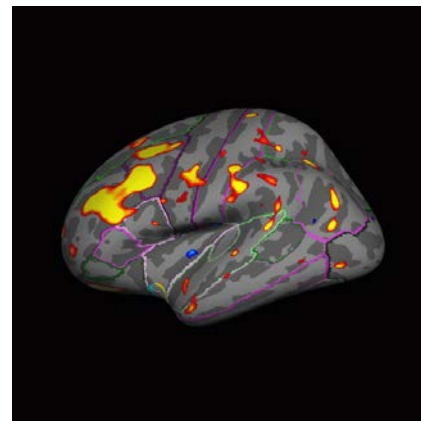
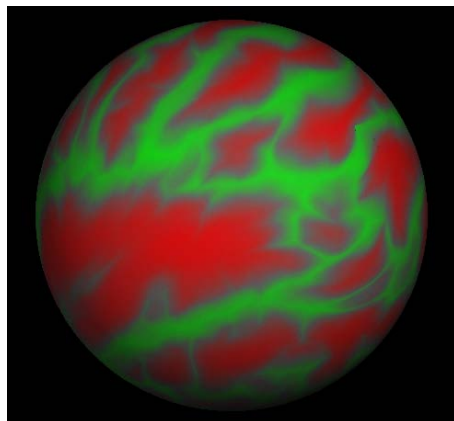
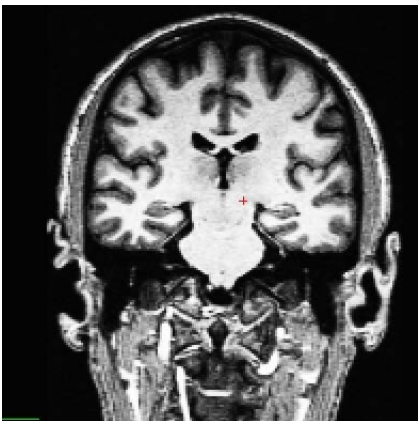
# Biomarkers Huntington's Disease (HD)

HD is a **genetic, neurodegenerative condition** causing loss of volumes and atrophy in the brain.

Symptoms include motor, cognitive and emotional disorder.

MRI as well as fMRI and DTI have a great potential in the design of new data-driven and user-specific biomarkers.

**Background**



# Biomarkers Huntington's Disease (HD)

MRI shows evidence of the disease **when it is already manifest**

**Background**

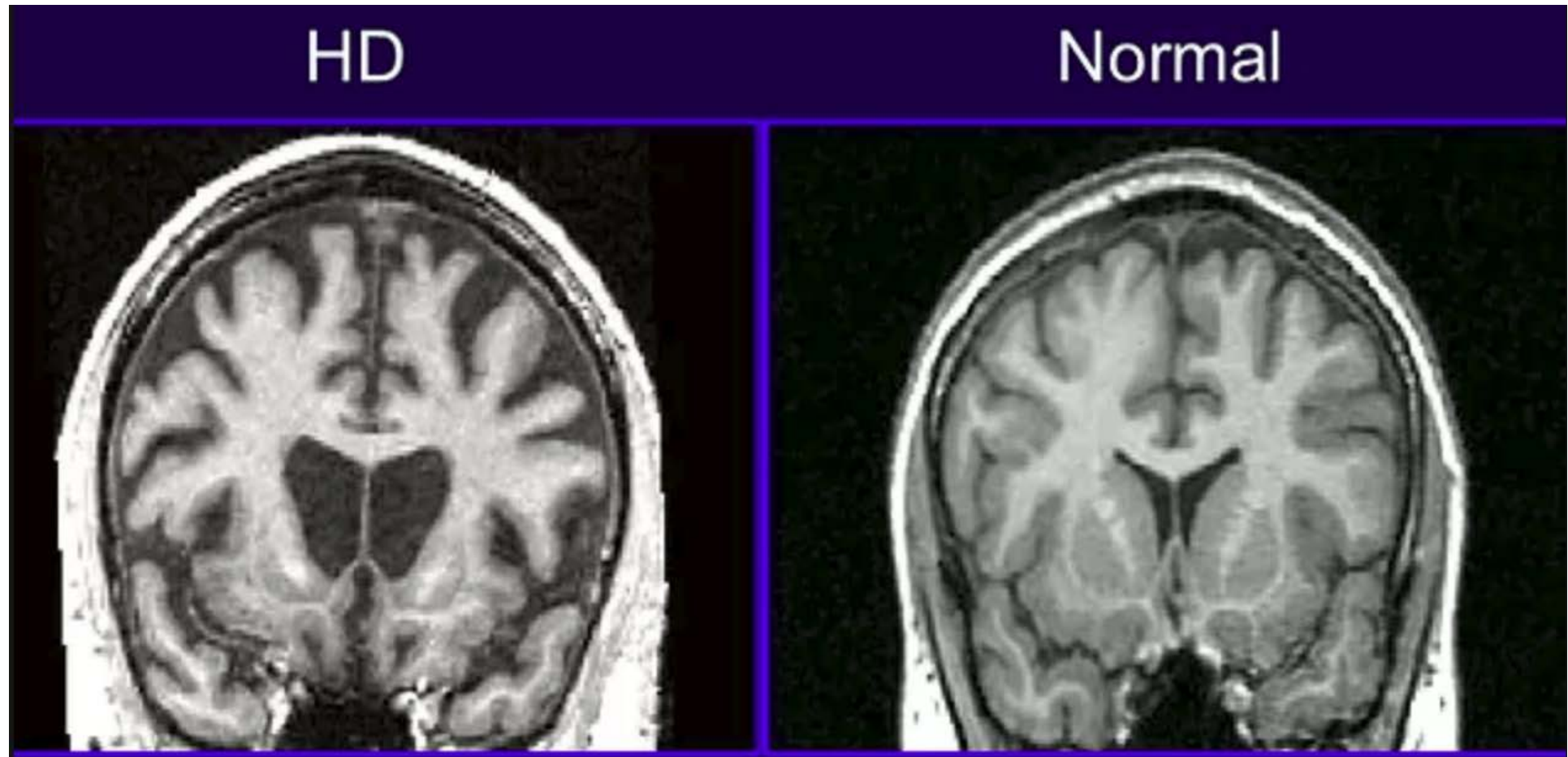
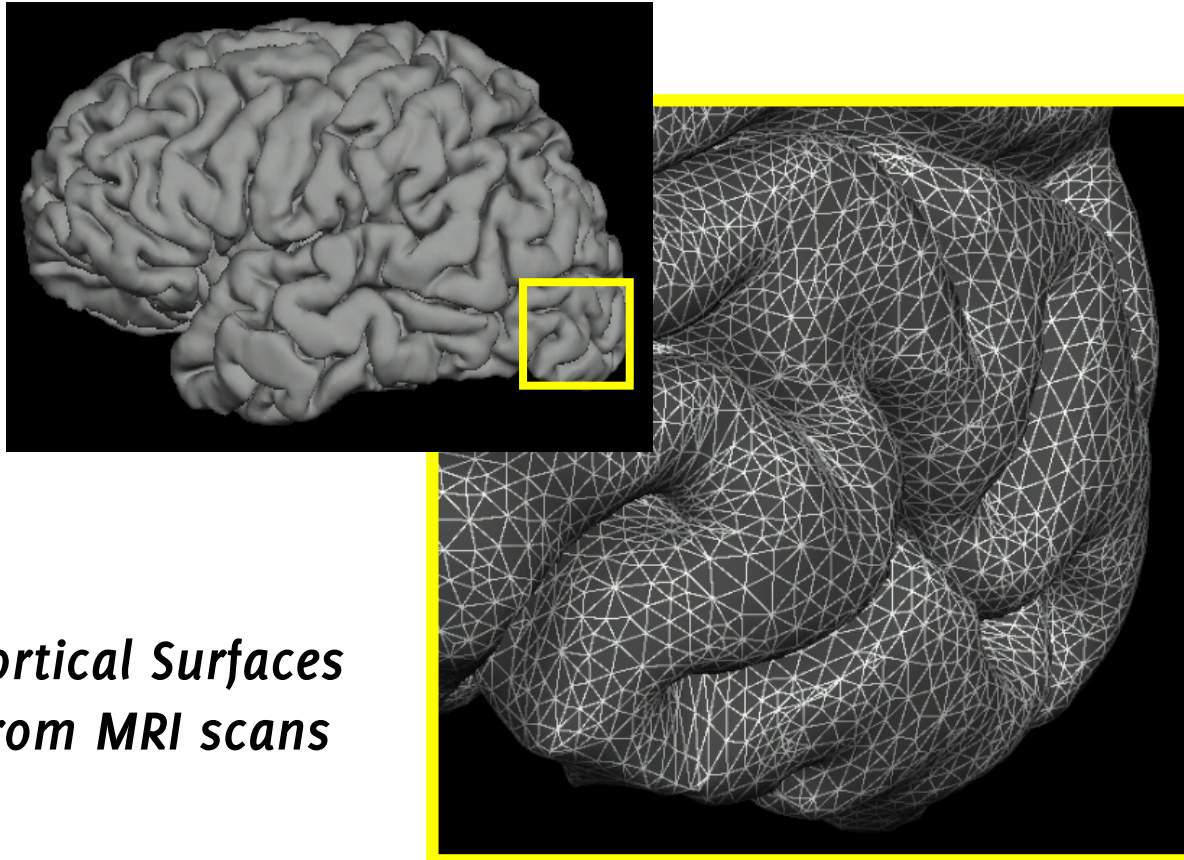


IMAGE CREDITS UC San Diego School of Medicine

# A data-driven biomarker for Huntington Disease (HD)

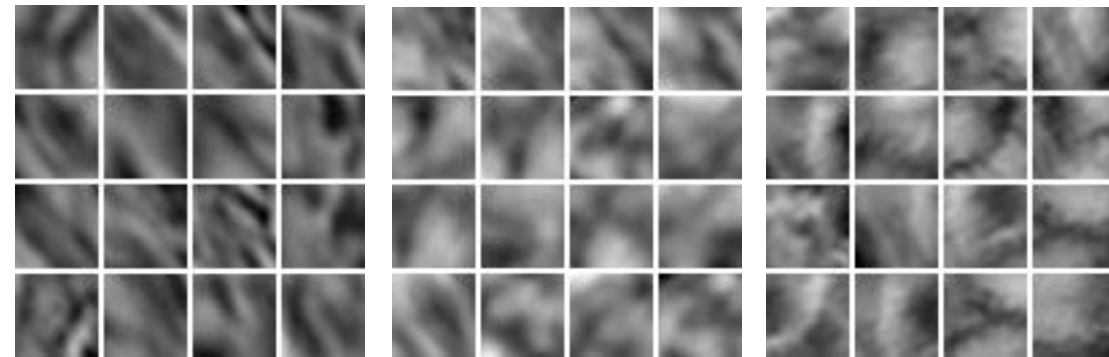
*We have developed a biomarker for quantitatively **assessing the progression of HD**. The biomarker **monitors changes in shape and thickness of cortical surfaces** by means of **user-specific and data-driven models**.*

## Background



*Cortical Surfaces  
from MRI scans*

*Atoms of dictionaries learned from  
brain regions*



Bankssts

Precuneus

Parsorbitalis

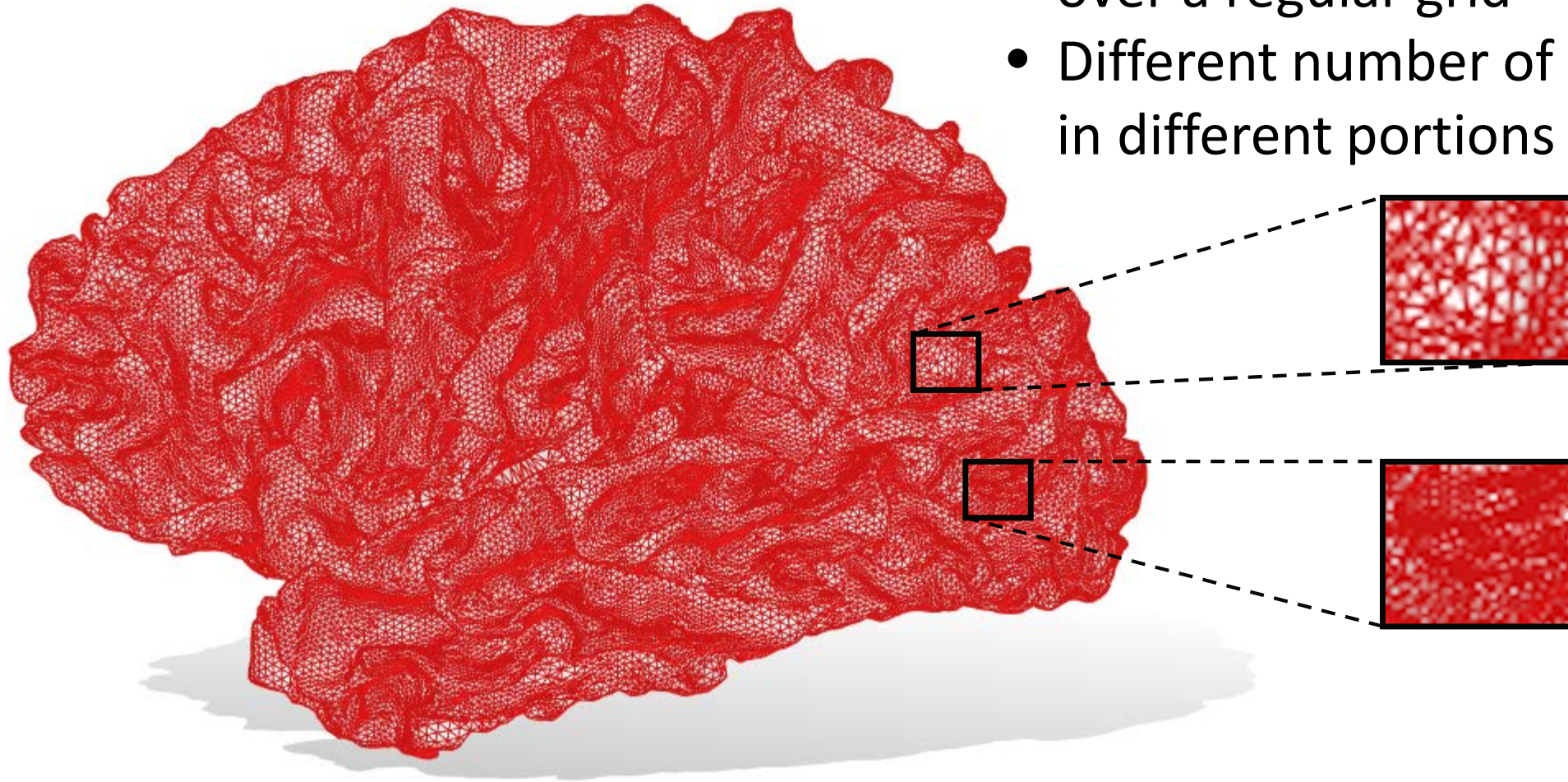
Collaboration with





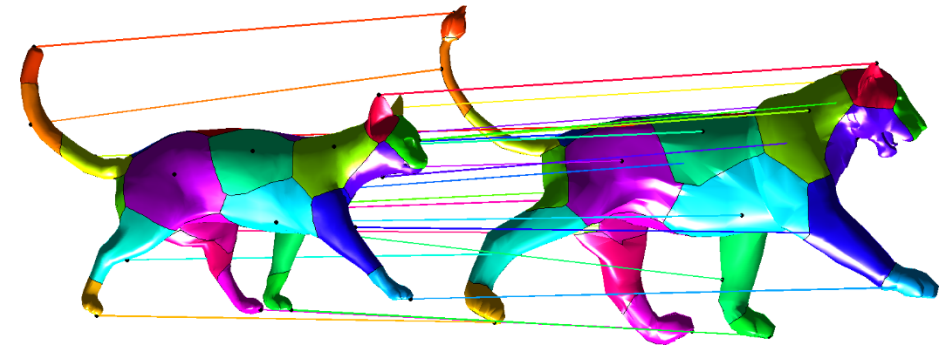
# Challenges of handling 3D surfaces

## Background



- Measurements are not set over a regular grid
- Different number of samples in different portions

# Thesis: Biomarker for HD



## Goals:

- Implement a biomarker that directly operates on 3D data points, without mapping these into 2D image representations
- Directly compare cortical surfaces of the same patient as deformable surfaces. Test whether differences/deformations are meaningful for predicting HD progression

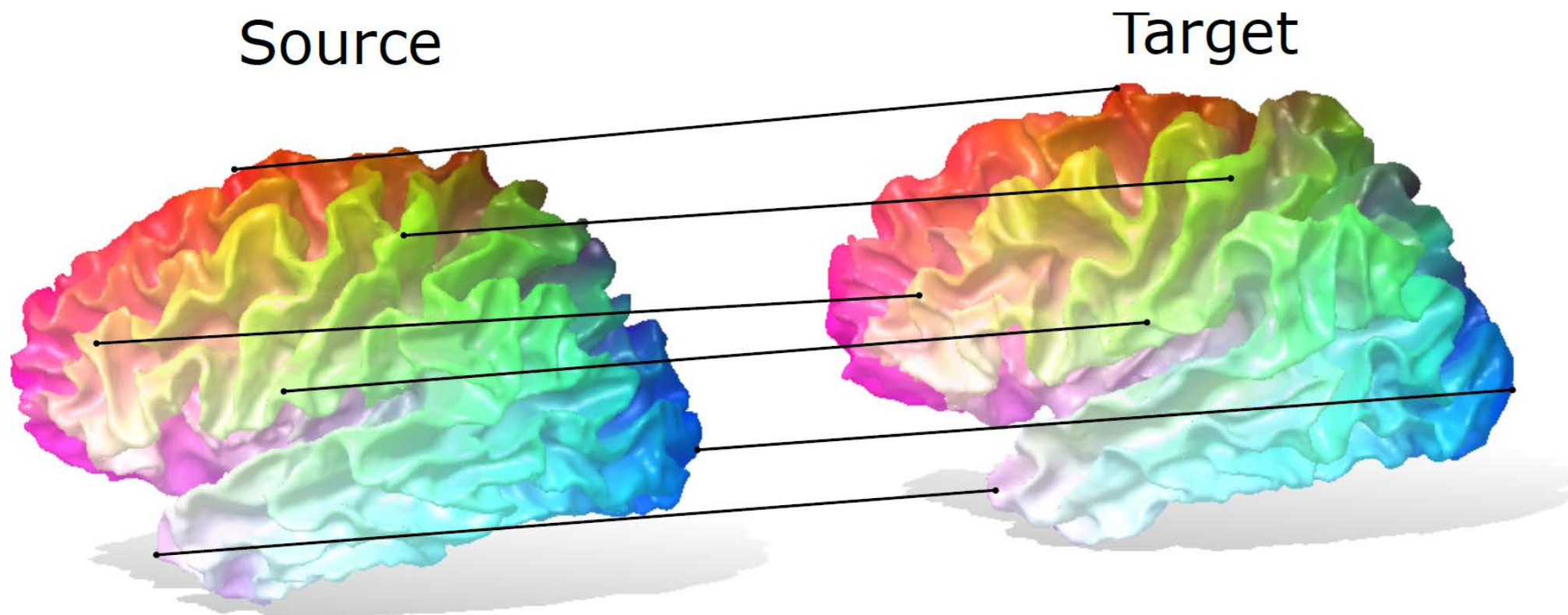
## Materials and Methods:

- Longitudinal scans of more than 80 HD patients and controls.
- Well defined testing procedures.
- Functional maps software for 3d shape registration



# Functional maps

An efficient and effective framework to estimate maps between 2D surfaces embedded in 3D.



# ML and CFD for diagnosing nasal pathologies

These thesis will be part of an established research collaboration with the aerospace department.

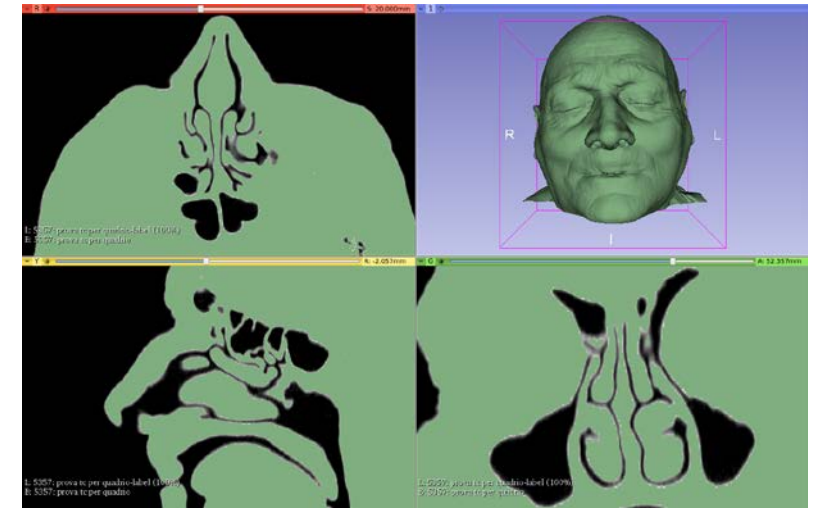
Goal: Design new machine learning methods that

- analyze CFD simulation of the human upper airways
- help surgeon to make the most informed choices

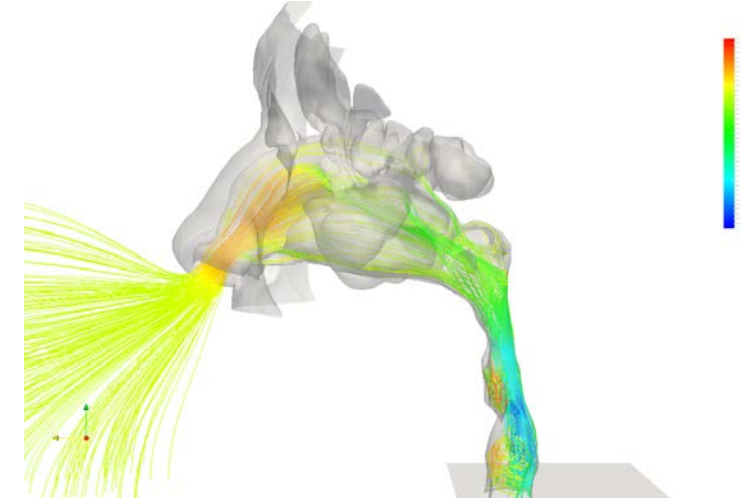
Why ML and CFD?

- Large failure rate of surgical corrections
- Lack of reliable diagnostic tools
- CT scan only gives information about shape of the nasal apparatus
- CFD can give information about the functional properties during breathing

*Output of a CT scan of the human upper airways*



*An illustration of CFD outcomes on a CT scan*

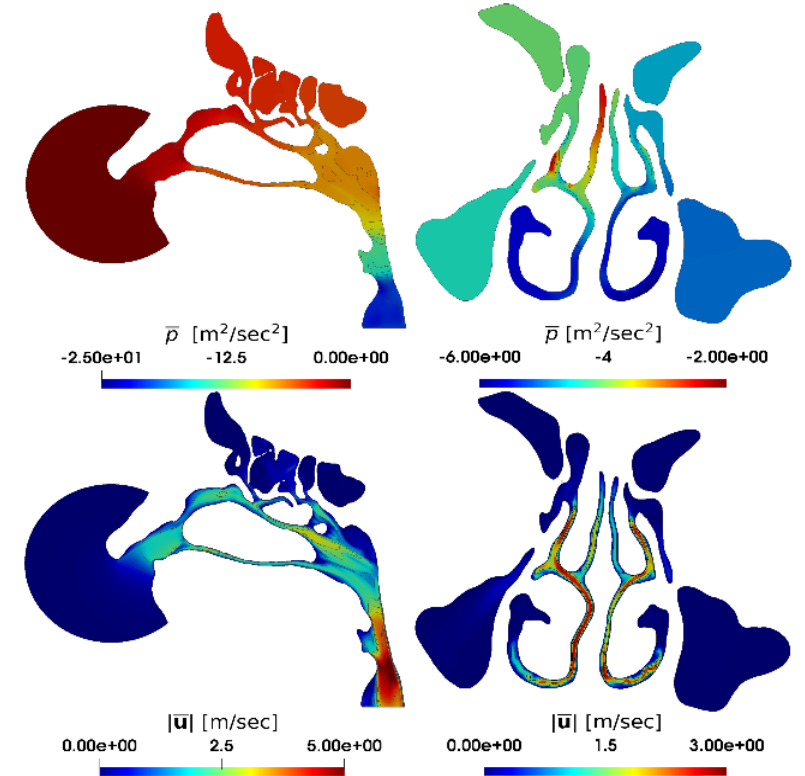


# ML and CFD for diagnosing nasal pathologies

## Challenges:

- **Dimensionality**, CFD's output is an high dimensional 3d (or 6d) -vector field, it has to be suitably transformed to be handled by ML models. On a disk it might take a few GB!
- **Registration**, CT scan provides very different shapes from different subjects which needs to be registered.
- **Very little supervision**, compared to the input dimension. Many different pathologies to diagnose

*Another illustration of CFD outcomes on a CT scan*



## Thesis Opportunities in ML and CFD

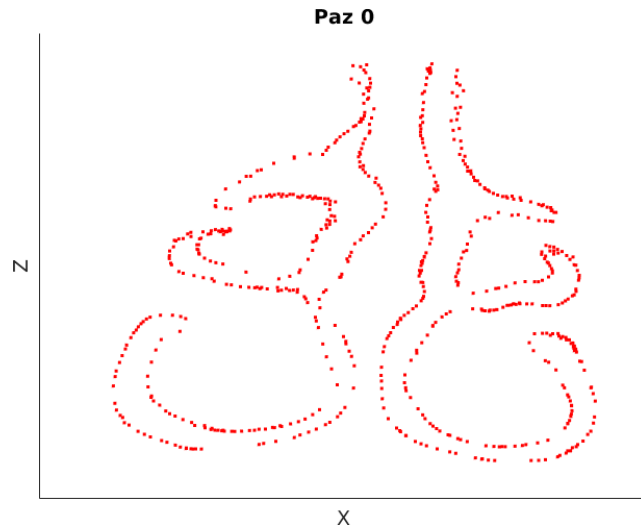
Research directions for multiple thesis:

- **Dimensionality reduction and feature design:** design the most representative features / **representation learning algorithms** to handle CFD output. Automatic selection of the most influential ones for medical diagnosis
- Investigate both grey-box and black-box approaches to perform dimensionality reduction on CFD.
- **CT registration by functional maps:** use these geometric that allows efficient inference and manipulation to “transport” and compare properties
- **Advanced functional maps:** formulate the most suitable constraints to our problem at hand.
- **Augmentation** design data-augmentation procedures to introduce realistic CT-scan manipulation to mimic pathological effects

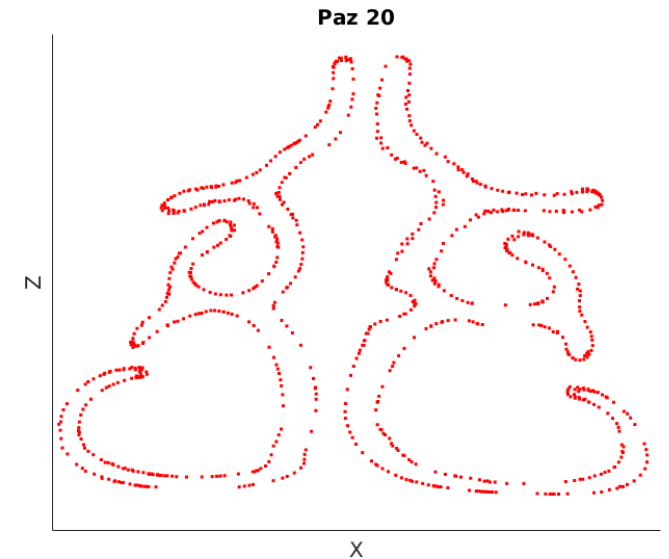
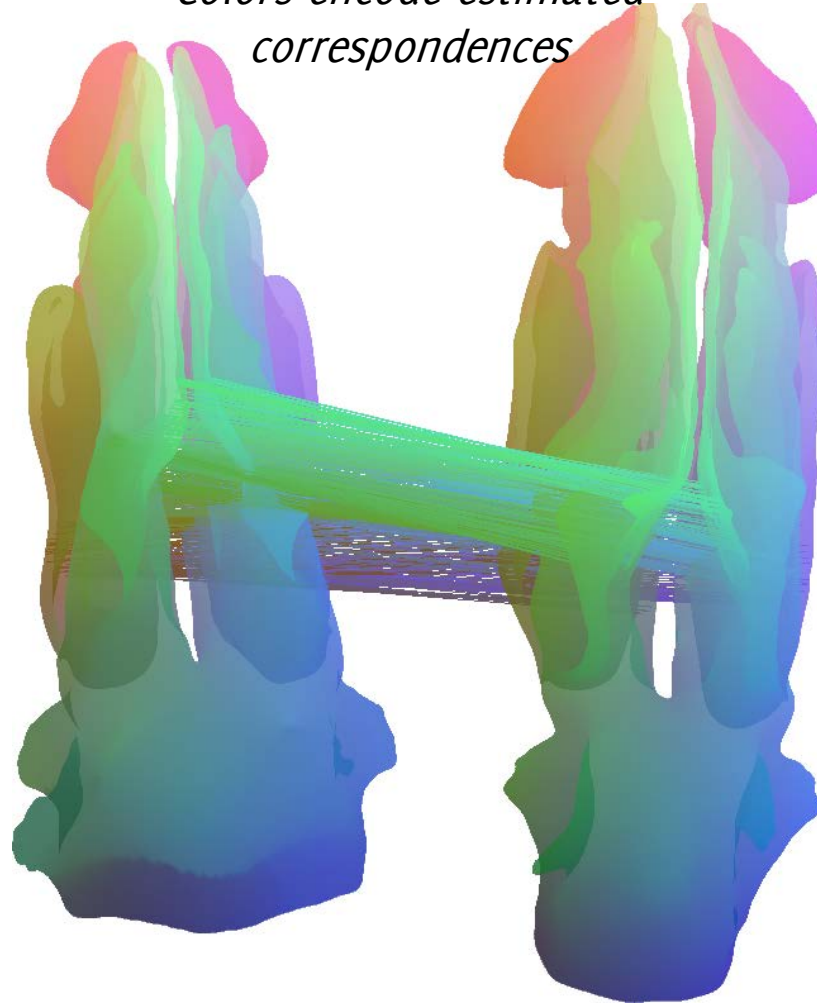
# Example of Registration via functional maps

*Noses from two subjects that are matched through functional maps.*

*Colors encode estimated correspondences*



*A reference section extracted from the first subject*



*A nasal section extracted in the second subject, estimated through the correspondences*



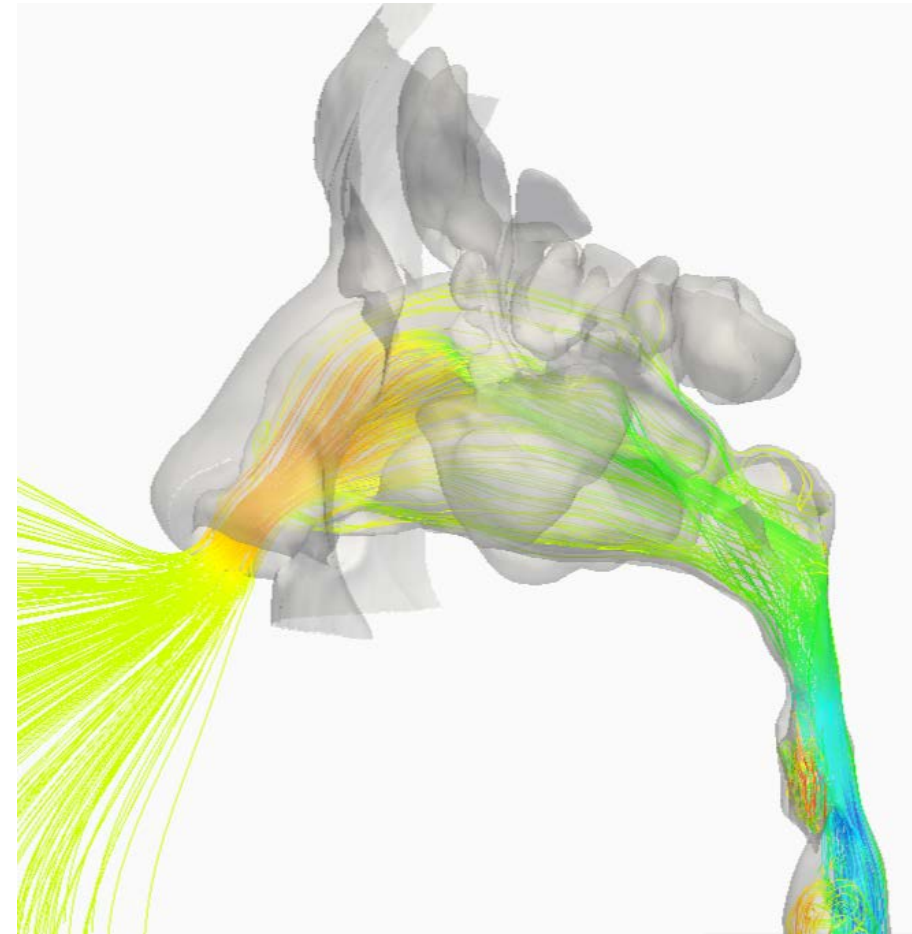
## Thesis Opportunities in ML and CFD

### Very Multidisciplinary environment:

This project brings together doctors, CFD expert, Data scientist, experts of geometric computer vision

### Methods and Materials:

- Features inspired to image analysis techniques.
- State-of-the-art feature selection methods from ML literature
- Many CFD simulations of patients provided with the medical diagnosis

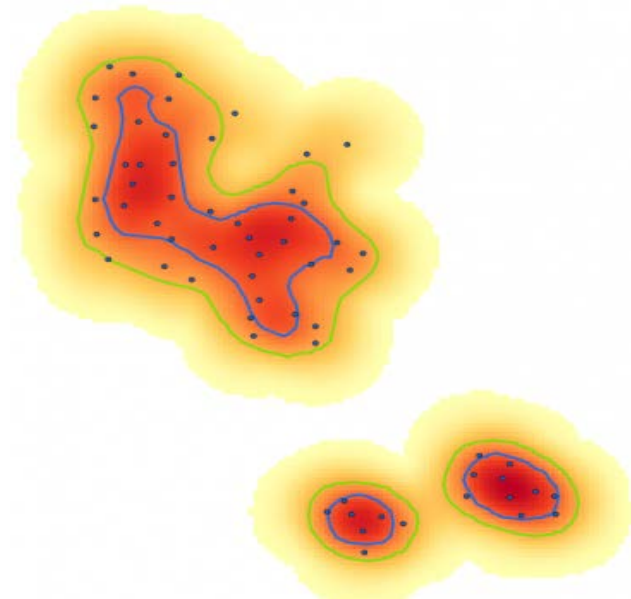
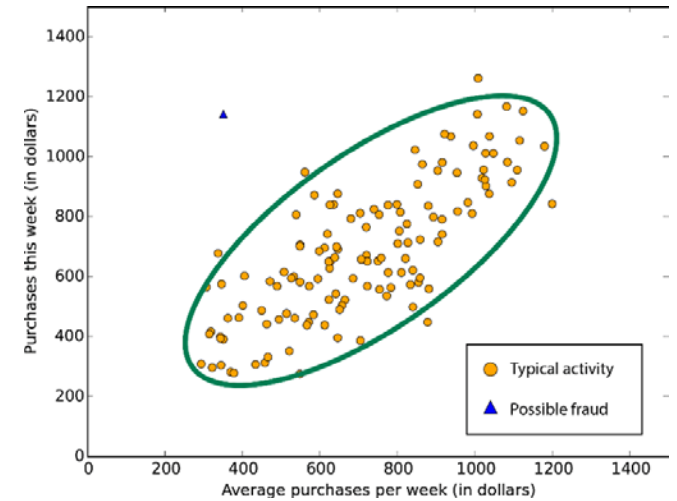


# Change/Anomaly Detection in Datastreams

## Anomaly detection in datastreams

Usually anomaly detection is carried out **modeling the density of normal data** and identifying **anomalies** as samples that fall in **low density regions**.

Non-parametric density models are much more flexible, but **extremely inefficient** (e.g. KDE). This is specially evident when the **data dimensionality** and the **number of samples** increases.

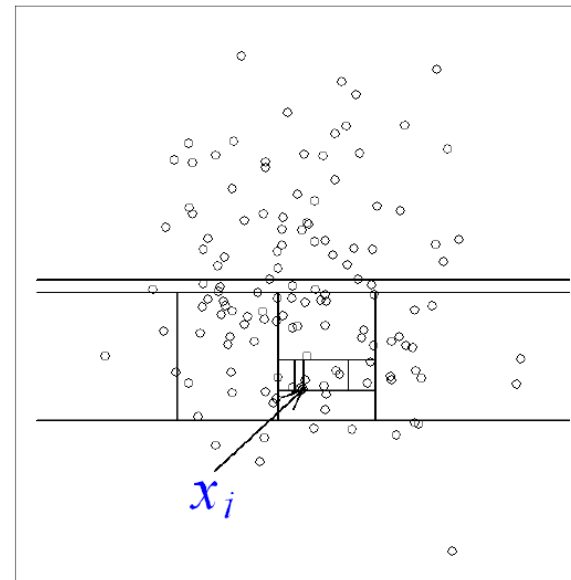
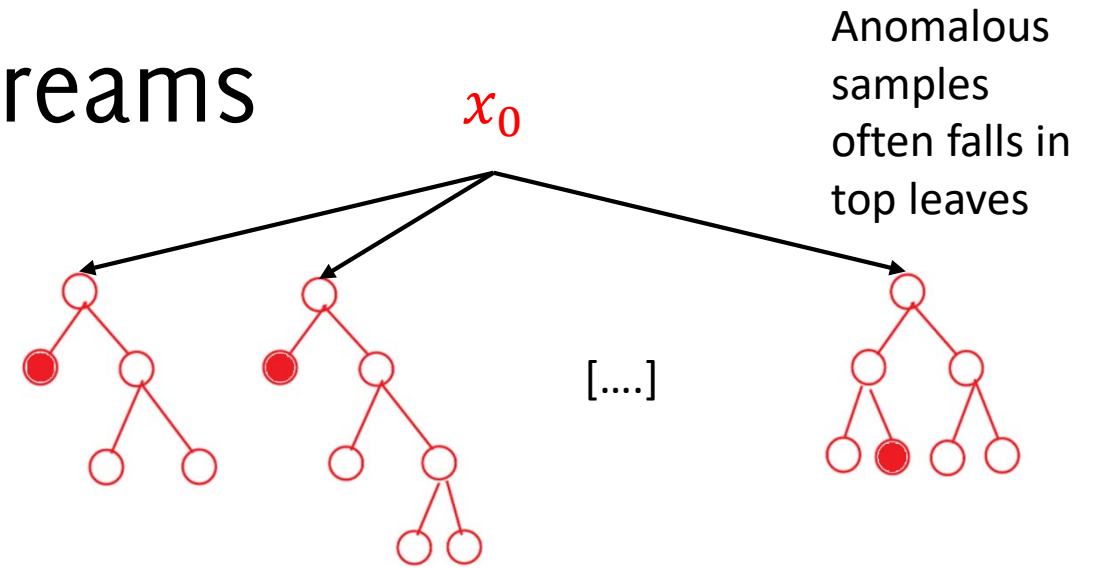
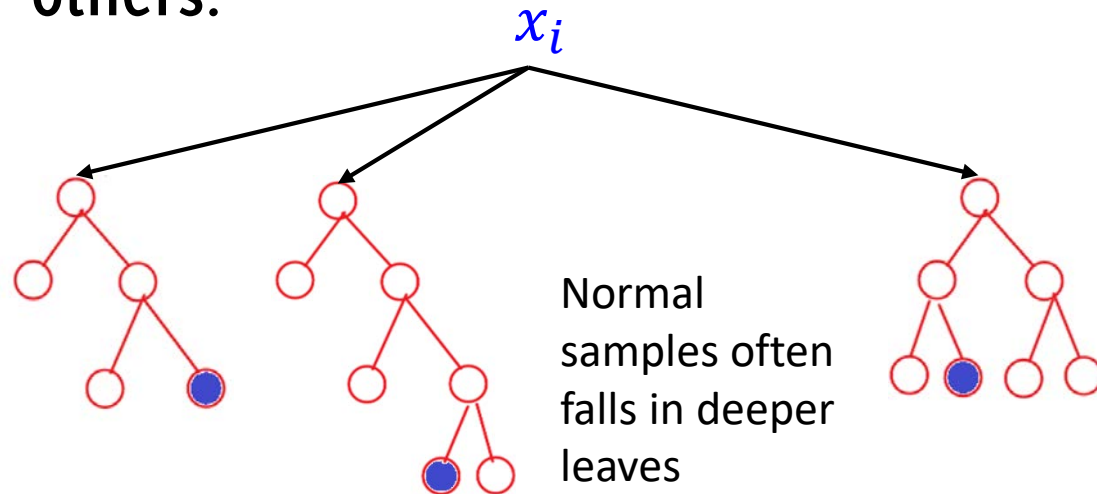


# Anomaly detection in datastreams

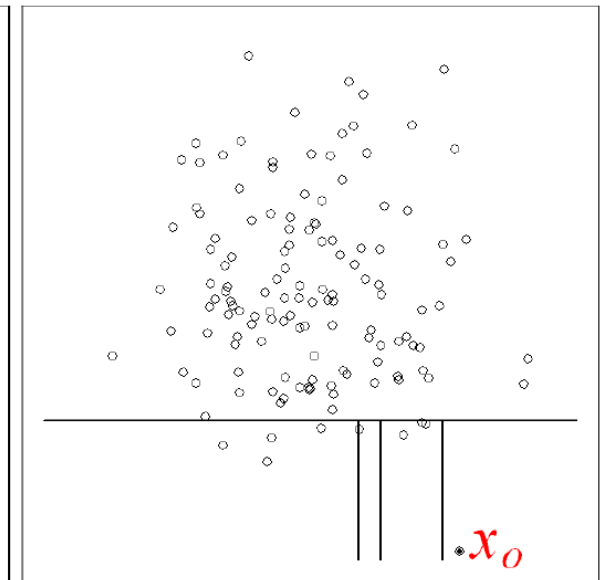
**Isolation Forest (IFOR):** Introduced a new paradigm in anomaly detection.

Instead of learning the density of normal data, IFOR models how **difficult** is to isolate each sample  $x$  from the others.

Background



(a) Isolating  $x_i$



(b) Isolating  $x_o$

## Thesis: Beyond IFOR

Goal: investigate improvements for IFOR based on:

- Design **novel decision rule** for IFOR, possibly based on **unsupervised loss**.
- Design **adaptation mechanisms** for IFOR to handle **evolving data-streams**.
- Include **expert-driven** information in the IFOR construction.
- **Optimized, parallel implementation** of IFOR on GPUs.
- Define **interpretability criteria** for the most apparent anomalies.
- Investigate extensions to **categorical data**.
- Adopt **copula theory** to factorize distributions.

Materials and Methods:

- Access to a server mounting GPUs will be provided.
- Dataset from our industrial partner.



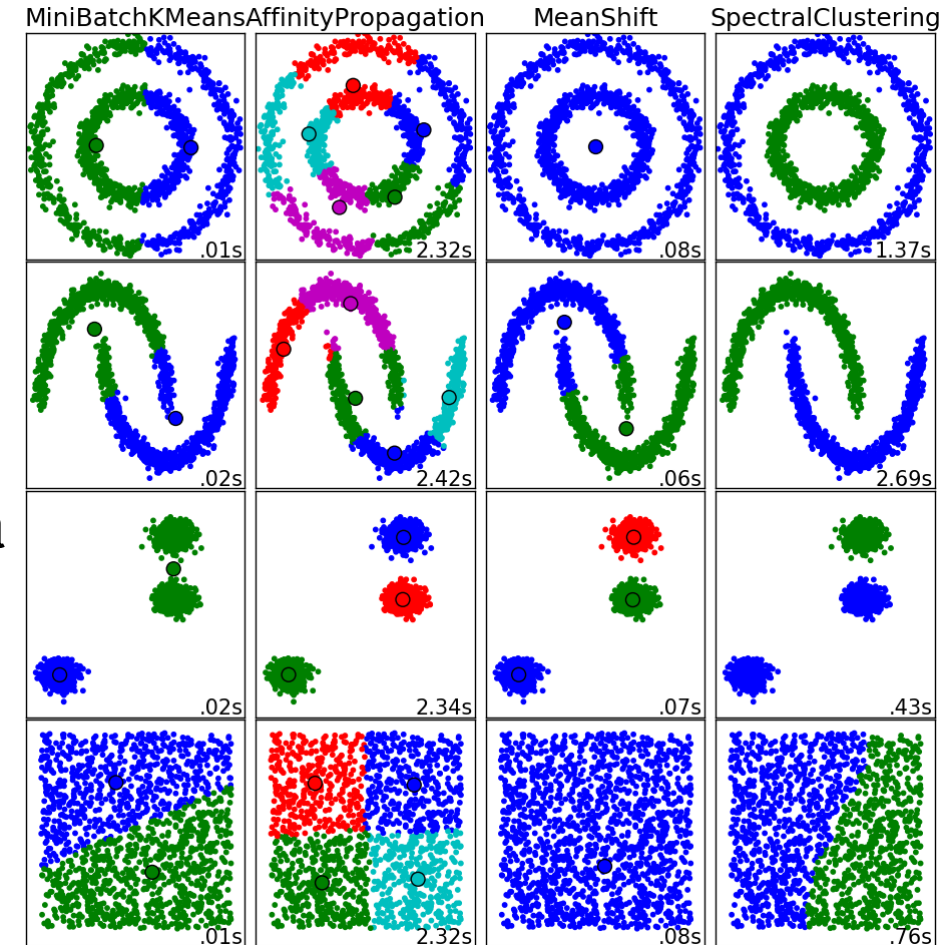
# Thesis: Isolation Based Clustering

**Goal:** exploit the potential of an isolation-based approach for clustering:

- Study the state-of-the-art of **clustering techniques**.
- Model the concept of “**isolation**” in IFOR.
- Design an innovative technique to **cluster data** taking advantage of “**isolation**”.

## Materials and Methods:

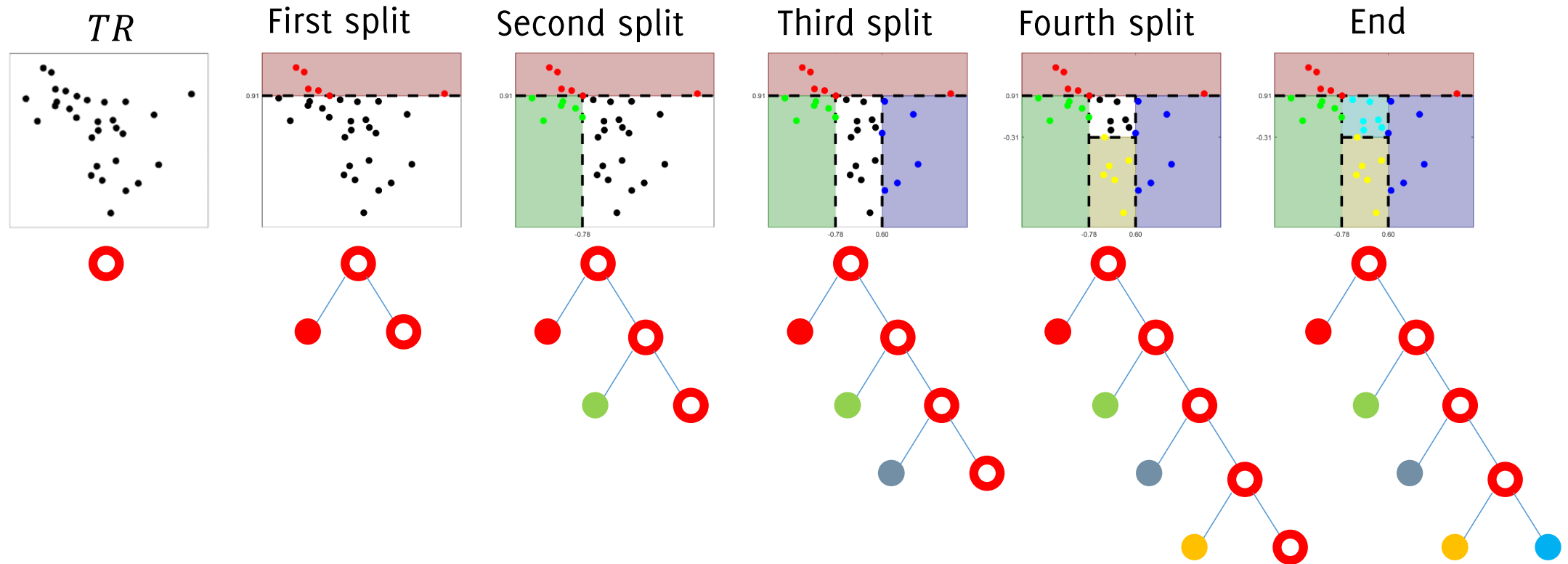
- Access to a server mounting GPUs will be provided.
- Dataset from our industrial partner.



# Change Detection in Datastreams

Efficient density models for detecting changes in high-dimensional datastreams, yet controlling false positives

Background



# Change Detection in Datastreams

## Goals:

- Design a **sequential monitoring scheme** (like CPM) **based on QuantTrees** to enable change detection in multivariate data streams
- **Study the probabilistic properties** of the QuantTree test statistics and use them to state parametric tests for change-detection purposes
- Investigate **incremental learning procedures** and their convergence guarantees for learning trees on very large training sets
- Investigate **new partitioning schemes** that are better suited for change-detection
- Design **parallel implementation** of an ensemble of histograms
- Design incremental/sequential procedures to construct QuantTrees

**Materials and Methods:** Experimental testbed on synthetic and real data, reference codes for QuantTrees to be used as a comparison.

# Sparsity and Convolutional Sparsity

Convolutional sparse models are a recent development of sparse representations

$$\mathbf{s} \approx \sum_{i=1}^n \mathbf{d}_i \circledast \boldsymbol{\alpha}_i, \quad \text{s. t. } \boldsymbol{\alpha}_i \text{ is sparse}$$

where a signal  $\mathbf{s}$  is **entirely encoded** as the sum of  $n$  convolutions between a filter  $\mathbf{d}_i$  and a coefficient map  $\boldsymbol{\alpha}_i$

Pros:

- Translation invariant representation
- Few small filters are typically required
- Filters exhibit very specific image structures
- Easy to use filters having different size

# Sparsity and Convolutional Sparsity

There are two major problems to be addressed when adopting models based on sparse-representations:

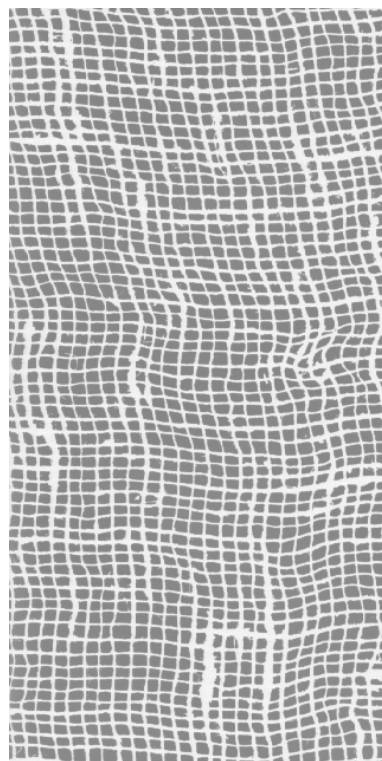
- Dictionary learning
- Sparse coding, namely computing the representation of an input signal w.r.t. the learned dictionary



# Thesis: Convolutional Sparsity

Design new sparse coding and dictionary learning schemes for improving image restoration performance

Training Image



Learned Filters

