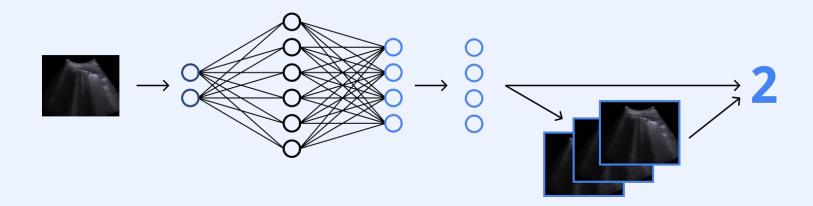
## LUS Images classification with uncertainty detection and image similarity

Medical Imaging Diagnostic

# Quick overview

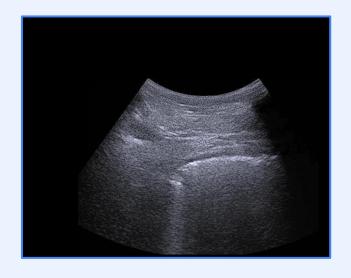
- 1. Frame into a classification model
- 2. Softmax output is analyzed by a classifier
- 3. If *false*, analyze the closer images score
- 4. Final score



# Briefly on data

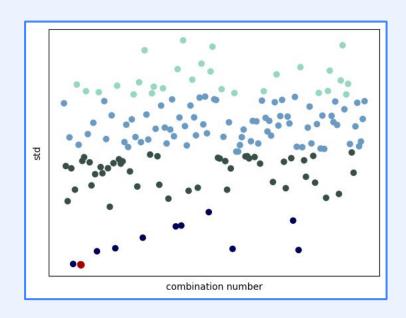
#### Lung Ultrasound Images

- 1. Images are scored from 0 to 3
- 2. This project works frame-wise
- 3. Augmentation was required
- 4. They seems in grayscale, but images are RGB



#### **Dividing patients**

To balance training, computed all possible 8-patients combinations stds between the number of frames per score -> took the smallest one



0	1	2	3
7888	7540	7189	7592

# First classifier

#### Choosing the model

- ResNet18
- VGG16
- SqueezeNet
- CNN from scratch

Added layers to fine-tune, but easy overfitting.
Working solution was to only add a layer to output the 4 classes

#### Augmentation

"Deep learning for classification and localization of covid-19 markers in point-of-care lung ultrasound"

E. Torri, R. Inchingolo, A. Smargiassi, G. Soldati, P. Rota, A. Passerini, R. J. G. van Sloun, E. Ricci, and L. Demi

- affine transformations
- multiplication with a constant
- Gaussian blurring
- horizontal flipping

### Confidence

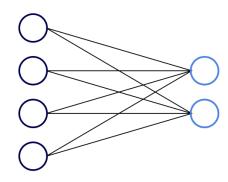
#### Threshold?

Independently from correct or wrong prediction and from the scores, the highest confidence in the softmax values were very similar

#### Understand the behaviour

Saving the softmax values and the prediction correctness

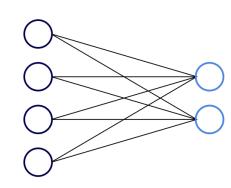
#### **Neural Network**



#### **Sigmoid** activation function

Basic model, any addition layer resulted in worse performance

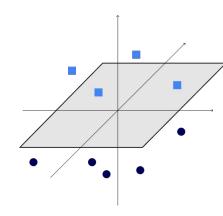
#### **Neural Network**



**Sigmoid** activation function

Basic model, any addition layer resulted in worse performance

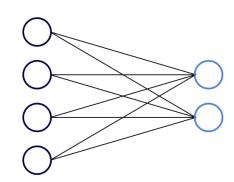
#### **SVC**



#### **Support Vector Classifier**

It required perfectly balanced dataset

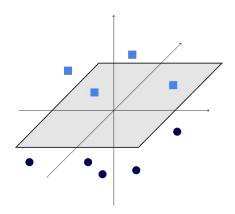
#### **Neural Network**



**Sigmoid** activation function

Basic model, any addition layer resulted in worse performance

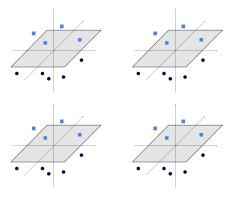
#### SVC



**Support Vector Classifier** 

It required perfectly balanced dataset

#### 4 SVCs



#### One for each class

The SVC used depended on the predicted score

# Similarity

#### Choosing the model

- ResNet18
- VGG16
- SqueezeNet
- CNN from scratch

Added layers to fine-tune, but easy overfitting. Working solution was to only add a layer to output the 4 classes

#### Hash

#### t-SNE

- similarity-wise (lighter, entire training set used)
- image-wise
- resnet18 embedding

First, this is not a classification method, is a...

Due to high time demanding, I chose the same number of images for each score for each training patients (and the same for testing).

Averaged the most X close images scores

For sim-wise ~Y%, for image-wise Z%

for a single frame, it required ~30 sec for each image

### Final results

#### Results

Only model

Only tsne

Model -> bin -> t/f -> tsne



### Extra

#### Example screen of similar images

It can also show the similar images, so in a "real world scenario", if there is uncertainty in a prediction, the X closest images can be plot:

### And that's it, thank you

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