



Normalize MRI for robust radiomics in oncology using Deep Learning

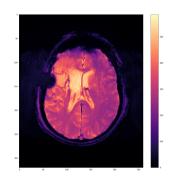
Sacha Dedeken

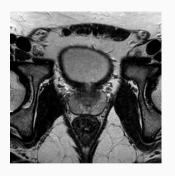
supervised by John Klein, Jérémie Boulanger from CRIStAL and David Pasquier. Alexandre Escande from Oscar Lambret

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Master 2 Data Science, University of Lille / Centrale Lille / IMT Nord-Europe

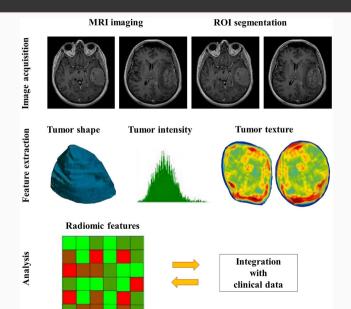
Human workflow of MRI





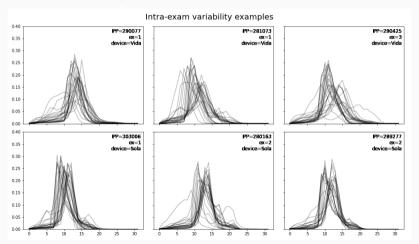
Common examination for diagnosis, prognosis and treatment decision for cancer

The radiomics revolution

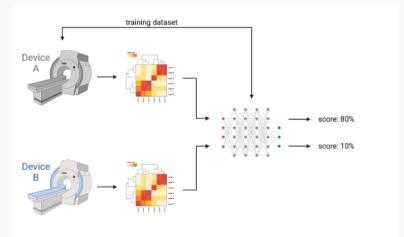


The robustness issue

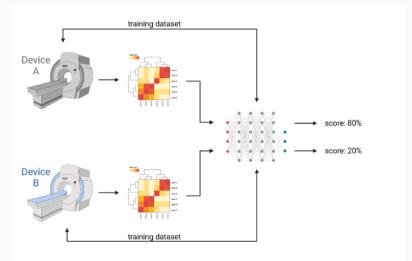
Three sources of variability : Electromagnetic field, Intra exam variability and Inter exam variability



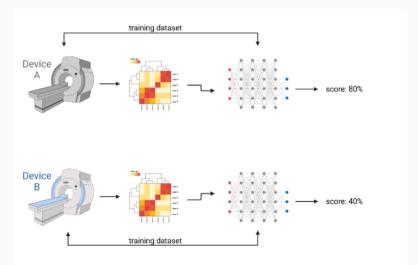
How to reduce inter-device variability?



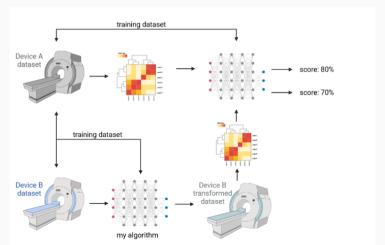
False good idea #1: Just merge the two dataset!



False good idea #2: Just train another algorithm!



New approach: Train deep learning to do histogram transport to harmonize the whole dataset



The most popular approach: ComBat

- Adapted from Johnson et al. (2007) in genomics
- This method works directly on radiomics that are described as:

$$y_{ij} = \alpha + X_{ij}\beta + \gamma_i + \delta_i \epsilon_{ij}$$

• It creates corrected values using maximum likelihood approach:

$$y_{ij}^{\text{ComBat}} = \frac{y_{ij} - \hat{\alpha} - X_{ij}\hat{\beta} - \hat{\gamma}_i}{\hat{\delta}_i} + \hat{\alpha}$$

Already validated for PET, CT and MRI with limited amelioration

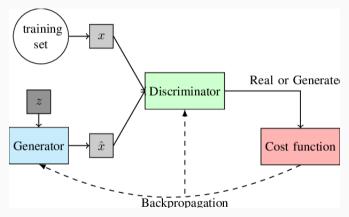
The good pre-processing (technical details)

- Create a mask to consider only brain voxels
- Extract histogram from Region Of Interest (ROI)
- Standardization (- mean and / by standard deviation)
- Normalization over 32 values + discretization

ID	exam	slice	device	h0	•••	h31
291755	2	18	1	0.02	•••	0.09

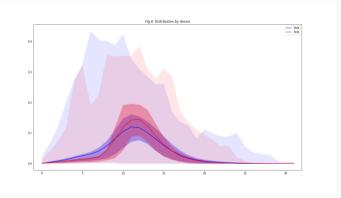
How to use GAN for histogram transport?

Generative Adversarial Network are perfect for this type of task!



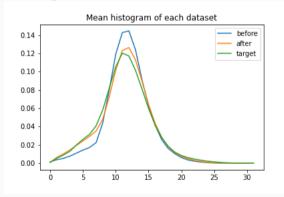
How to evaluate GAN performances

Visual assessment comparing to the whole distribution is very limited: it's hard to look after a correct distribution movement!



How to evaluate GAN performances

Moments convergence (mean and diagonal of covariance matrix)



$$score = \frac{\chi^{2}(h_{s}, h_{t}) - \chi^{2}(h_{g}, h_{t})}{\chi^{2}(h_{s}, h_{t})}$$

- +60% improvement on mean test score, but...
- not sensitive to individual tranformations
- more a sanity check than a satisfying metric

We can do better: an 8-steps experimental protocol

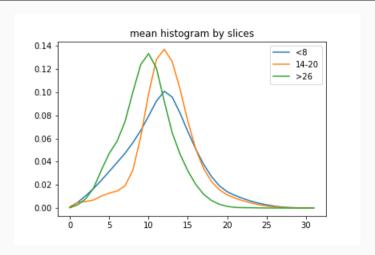
- Dataset analysis on size, modality, body region, prediction possibility
- 2. Pre-processing
- 3. Train-test split both devices
- 4. Algorithm training $Gen:Vida(Vida_{train} + Vida_{test})$ $Dis:Skyra_{train}$

- 5. **Histogram generation** Gen: $Vida_{train}$, $Vida_{test}$ $\implies Vida_{train}^G$, $Vida_{test}^G$
- 6. Radiomics creation
- 7. Radiomics-based predictor evaluated on 6 conditions
- 8. **Results analysis** with comparison to ComBat

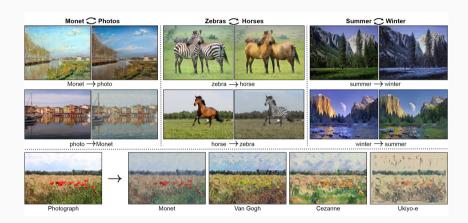
Personal view

- Very happy to have pursued this project in M2
- Finding correct medical data and clean them takes time
- This literature review, experimental protocol and python toolbox pave the road to future projects
- Now fully convinced that I want to work in data science and machine learning applied to the medical field
- Found a directly linked internship in the Radiomics Research Team of SOPHiA Genetics near Bordeaux
- I consider doing a thesis in deep learning for radiomics

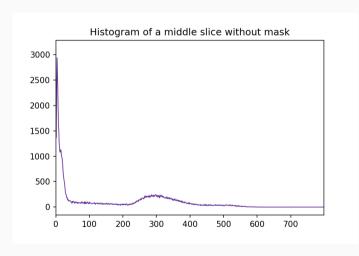
Bonus #1: ConditionalGAN



Bonus #2: CycleGAN



Bonus #3: Brain Extraction Tool



Bonus #4: An excellent paper very similar to our approach

comBat versus cycleGAN for multi-center MR images harmonization

Stenzel Cackowski¹

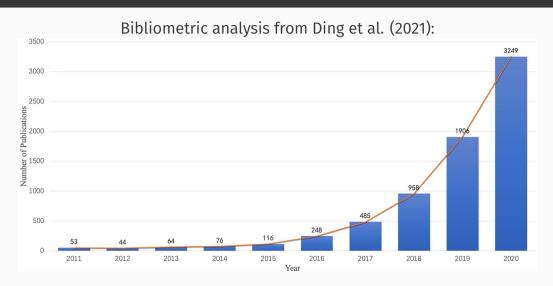
STENZEL.CACKOWSKI@UNIV-GRENOBLE-ALPES.FR

Emmanuel L. Barbier¹ Michel Dojat ¹ Thomas Christen ¹ EMMANUEL.BARBIER@UNIV-GRENOBLE-ALPES.FR MICHEL.DOJAT@UNIV-GRENOBLE-ALPES.FR THOMAS.CHRISTEN@UNIV-GRENOBLE-ALPES.FR

¹ Université Grenoble Alpes, Inserm U1216, Grenoble Institut for Neurosciences

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Bonus #5: Radiomics between revolution and buzz word



Bonus #6: Evaluation on 6 subgroups

Train your supervised predictor on $Skyra_{train}$ then evaluate it on 6 conditions to answer the following question "Does the classifier obtain a good score...":

- *Skyra_{train}*: ...on its training data?
- Skyra_{test}: ...on new data from the same device?
- $Vida_{train} + Vida_{test}$: ...on new data from another device without harmonization?
- $Vida^G_{train}$: ...on new but harmonized data, coming from the GAN training set?
- $Vida_{test}^G$: ...on new but harmonized data, unseen by the GAN?
- Vida_{combat}: ...on new but harmonized data via the ComBat method?