

T8 - AI & Big Data

T-DEV-810

Image recognition

Project

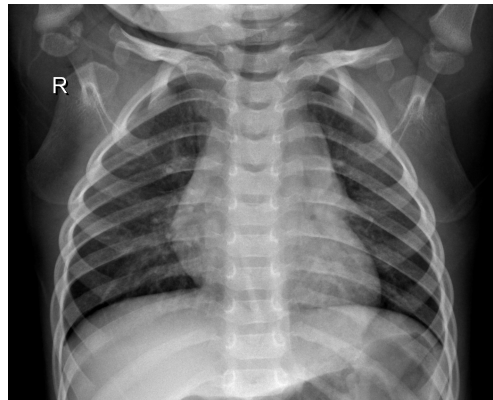


Image recognition

binary name: imageRecognition_{\$AcademicYear_}\$GroupNumber.zip
language: whatever works

Given some X-ray images, use machine learning to help doctors detecting pneumonia.



Doctors granted you access to 3 datasets if needed.



It's up to you to decide when and how to use the dataset (training, testing, evaluating performance, tuning parameters, ...).

You must:

- use a train-validation-test procedure,
- use a cross validation procedure,
- compare your results with a simple train test split,
- use one of the datasets to tune your algorithms.

Specialists insist on the importance you to discover/explore/test various methods, **and to compare results**.



optimization, feature engineering, metrics, PCA



A clear and concise way to present results should always prevail.

You are expected to deliver:

- technical documents
a **Jupyter notebook-like file**, containing code and text, possibly graphics and an **html-file** to prove your results without rerunning the code
- a synthesis document
a **pdf file** to sum up your results and figures



There are ways to save a trained algorithm and load it afterwards in order to obtain the same results when you run it again!

BONUS

You can improve your project in many ways, including:

- implementing a self-organizing map to help you visualize your result,
- learning deeply through neural networks,
- predicting on 3 classes: no pneumonia, virus pneumonia, bacteria pneumonia,
- ...

RECOMMENDATIONS

- Time and space

So much effort for such a poor result.

P. Sorentino

Think carefully about needed resources (time and space) and a procedure, **before starting implementation**. For quicker or better results, you may *-or may not-* want to transform your data.



Some libraries could help you.

- Bad habits

Do not let your algorithm(s) take bad habits.

Find the correct balance between **bias** and **variance**.

Cross validation procedures can help you solve this problem.

It's also a good idea when you don't have a lot of data.

Algorithms' parameters can also help you to find a trade-off between bias and variance, this is why you need a good understanding of how they work.

- Good metrics

Find an explicit way to show your result, in a readable way.

You must choose among many various metrics and **select the appropriate one(s)**.

Look at more advanced metrics like ROC-AUC score.

It will give you a deeper understanding of your results.

You should also be able to explain what AUC measures and its advantages over other metrics.



First ML project!? You should dig a bit and find answers: the only way out is the way in!