

DEEPHEALTH

Deep-Learning and HPC to Boost Biomedical Applications for Health

Pipelines for Medical Imaging Use Cases & Requirements for Benchmarking

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Evaluation schemes for Big data and Al Performance of high Business impact

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About DeepHealth



- Put HPC computing power at the service of biomedical applications with DL needs and apply DL techniques on large and complex image biomedical datasets to support new and more efficient ways of diagnosis, monitoring and treatment of diseases.
- Facilitate the daily work and increase the productivity of medical personnel and IT professionals in terms of image processing and the use and training of predictive models without the need of combining numerous tools.
- Offer a unified framework adapted to exploit underlying heterogeneous HPC and Cloud architectures supporting state-of-the-art and next-generation Deep Learning (AI) and Computer Vision algorithms to enhance European-based medical software platforms.

Key facts



Duration: 36 months

Starting date: Jan 2019



Budget

14.642.366 € EU funding 12.774.824 €



22 partners from 9 countries:

Research centers, Health organizations, large industries and SMEs

Research Organisations















Health Organisations















SMEs



















Developments & Expected Results



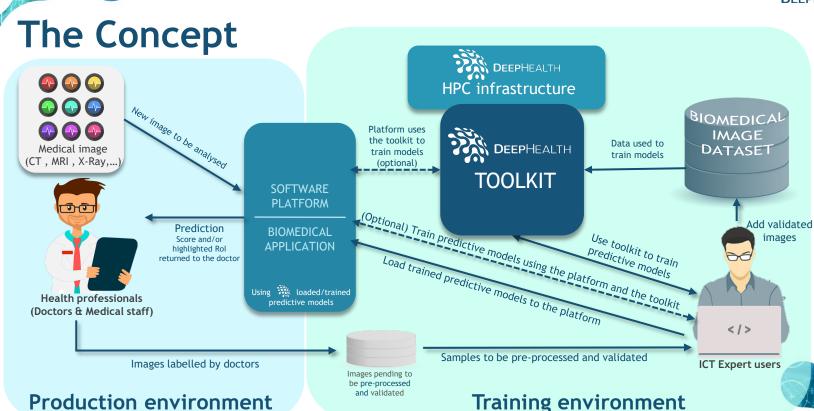
- The DeepHealth toolkit
 - Free and open-source software: 2 libraries + front-end.
 - EDDLL: The European Distributed Deep Learning Library
 - ECVL: the European Computer Vision Library



- Ready to run algorithms on Hybrid HPC + Cloud architectures with heterogeneous hardware (Distributed versions of the training algorithms)
- Ready to be integrated into end-user software platforms or applications
- **HPC infrastructure** for an efficient execution of the training algorithms which are computationally intensive by making use of heterogeneous hardware in a transparent way
- Seven enhanced biomedical and AI software platforms provided by EVERIS, PHILIPS, THALES, UNITO, WINGS, CRS4 and CEA that integrate the DeepHealth libraries to improve their potential
- Proposal for a structure for anonymised and pseudonymised data lakes
- Validation in 14 use cases (neurological diseases, tumor detection and early cancer prediction, digital
 pathology and automated image annotation).





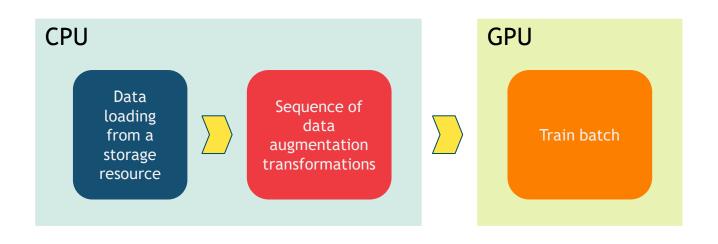








A **pipeline** is a set of operations **sequentially applied** to a data block (subset of samples)









Data Pipeline AI/ML/DL Model Pipeline **Training Pipeline** Persistent AI/ML/DL AI/ML/DL Dataset Data Data Data Data Solution Model Model Data Deployment design Acquisition Curation **Partitioning** Augmentation training evaluation Storage

Training Pipeline is at the core of the **Model Pipeline** which in turn is considered part of the **Data Pipeline**

Both pipelines are suitable for business and research applications

The whole *Data Pipeline* is applicable to any sector. Our project is focused on the Health sector







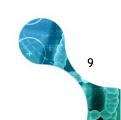


Data & Model Pipelines - Dataset Design

AI/ML/DL AI/ML/DL Persistent Data Data Data Solution Dataset Data Model Model Data **Deployment** design Acquisition Curation **Partitioning** Augmentation Storage training evaluation



- Data types and formats -identification and definition
- Metadata definition
- Data Lake structure definition
- Guidelines / HOW-TOs
- Etc.





Data & Model Pipelines - Acquisition + Curation + Storage

AI/ML/DL Persistent AI/ML/DL Data Dataset Data Data Data Solution Model Model Data Curation design Acquisition **Partitioning** Deployment Augmentation training evaluation Storage



- Data acquisition continuum
- Data cleansing/cleaning/wrangling/crunching
- Aggregated data/values computation
- Implementation of the data-lake definition
- Creation of users and permissions or make data public









Dataset design

Data Acquisition Data Curation Persistent Data Storage

Data Partitioning Data Augmentation AI/ML/DL Model training AI/ML/DL Model evaluation

Solution Deployment

Model training loop:

- Partitioning in training/validation/test subsets
- Data augmentation on-the-fly
- Training and evaluation of models
- Cloud & High Performance Computing requirements





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Skin Lesion Detection and Classification



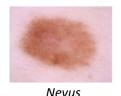
- Use case n° 12 of the DeepHealth project is based on the International Skin Imaging Collaboration dataset
- Aims: identification (segmentation) and diagnosis (classification) of skin lesion images among different classes







Carcinoma







Dataset design

Data Acquisition

Data
Curation
Persistent
Data
Storage

Retrospective acquisition

- 23.906 annotated images
- Publicly available on the ISIC archive website
- jpeg data format

Data Partitioning

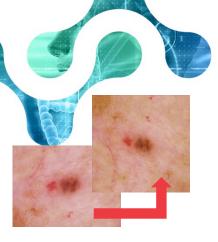
Data Augmentation

Model training

Model evaluation

- Training 19.000
- Validation 906
- Test 4.000





Skin Lesion Detection and Classification



Performed using the DeepHealth toolkit. Models are already available in the front-end.



Data Augmentation AI/ML/DL Model training

Explainability plays a fundamental role in this context. Ensuring Confidence Calibration and providing a Visual Explanation of the models is essential to support clinicians.



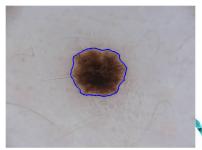






AI/ML/DL Model evaluation

Jaccard Index (Intersection over Union)



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Needs & Requirements

Evaluate datasets in terms of

- 1. Findability where should a data scientist search for the dataset?
- 2. Availability how long does a data scientist need to start the initial exploratory data analysis?
- 3. Interoperability how long does a data scientist need to start training AI/ML/DL models with a dataset?
- **4. Reusability** are previously obtained results with a dataset public and available to other researchers / data scientists?
- 5. **Privacy** / **Anonymisation** can the dataset be made public without compromising the identity of individuals?
- **6. Quality** is the dataset biased or unbalanced? What procedure has been followed to validate annotations?





Needs & Requirements

Evaluate Deep Learning libraries in terms of

- Speed-up is distributed learning really efficient?
- 2. Convergence does the distributed learning reach the same model accuracy in less time?
- **3. Usability** how long does a developer need to use the libraries effectively?
- 4. Integrability how difficult is it to integrate the libraries as part of solutions to deploy?
- **5. KPIs:** time-of-training-models (**totm**), performance/power/accuracy trade-off, etc.
- **6.** Others can you help us to evaluate other aspects?





Needs & Requirements

Evaluate Software Platforms in terms of

- 1. **Usability** how long does a domain application expert need to manage the software tool effectively?
- 2. Completeness does the application platform provide all the algorithms/procedures/functions to allow domain application experts to easily define the sequences of steps to implement the data and/or model pipelines?
- **3. Compatibility** how many data formats does the platform admits to import/export data and models from/to other frameworks?
- 4. **KPIs:** time-to-model-in-production (**ttmip**), time-of-pre-processing-images (**toppi**), etc.
- **5.** Others can you help us to evaluate other aspects?





