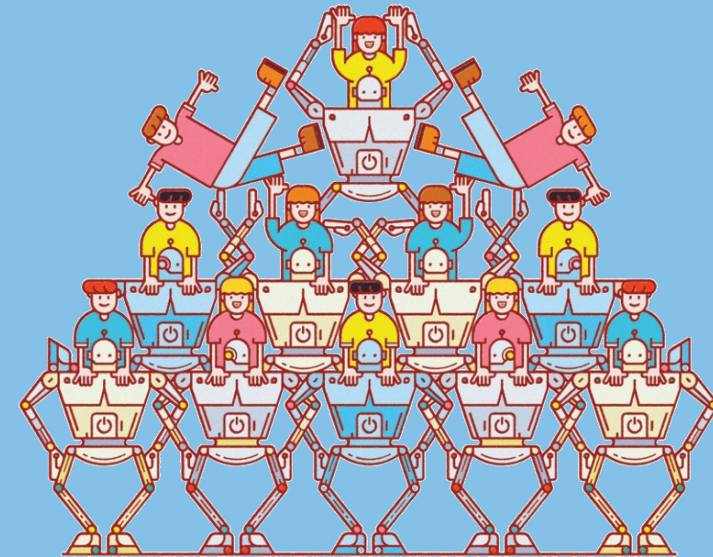


WP1: Human-AI interaction Kickoff meeting

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LISA, ULB



Supported by Service public de Wallonie – Recherche under grant n°2010235 « ARIAC BY DIGITALWALLONIA4.AI »

WP1 kickoff meeting: goals

Get to know each other and start building a network of the TRAIL researchers working on topics related to human-AI interaction

Foster collaborations among researchers, in particular PhD students, working on similar topics.

WP1 kickoff meeting: program

14h: Alberto Franzin: presentation of WP1

14h10: Adrien Foucart: DL for histopathology with multiple annotations

14h40: Contributed presentations (10m each): Elias Fernández, Antonio García-Díaz, Romain Mormont, Alberto Franzin

15h20: Question time for contributed presentations, with coffee

15h50: Contributed presentations (3m each + questions): Charlotte Nachtegaele, Anaïs Halin, Natarajan Chidambaram, Laura Gálvez Jiménez, Alexandre Englebert

16h25: Christine Decaestecker: bias in AI

17h: Networking and informal discussions

Human-AI interaction: what, why, how

What: Establish and ease relations and interactions between AI and humans (experts or users)

Why: Improve AI performance and open up realistic perspectives on the application of AI in sensitive areas (medicine, justice, etc.)

How: 5 tasks

Human-AI interaction: an example

- Supervised learning makes use of labeled data

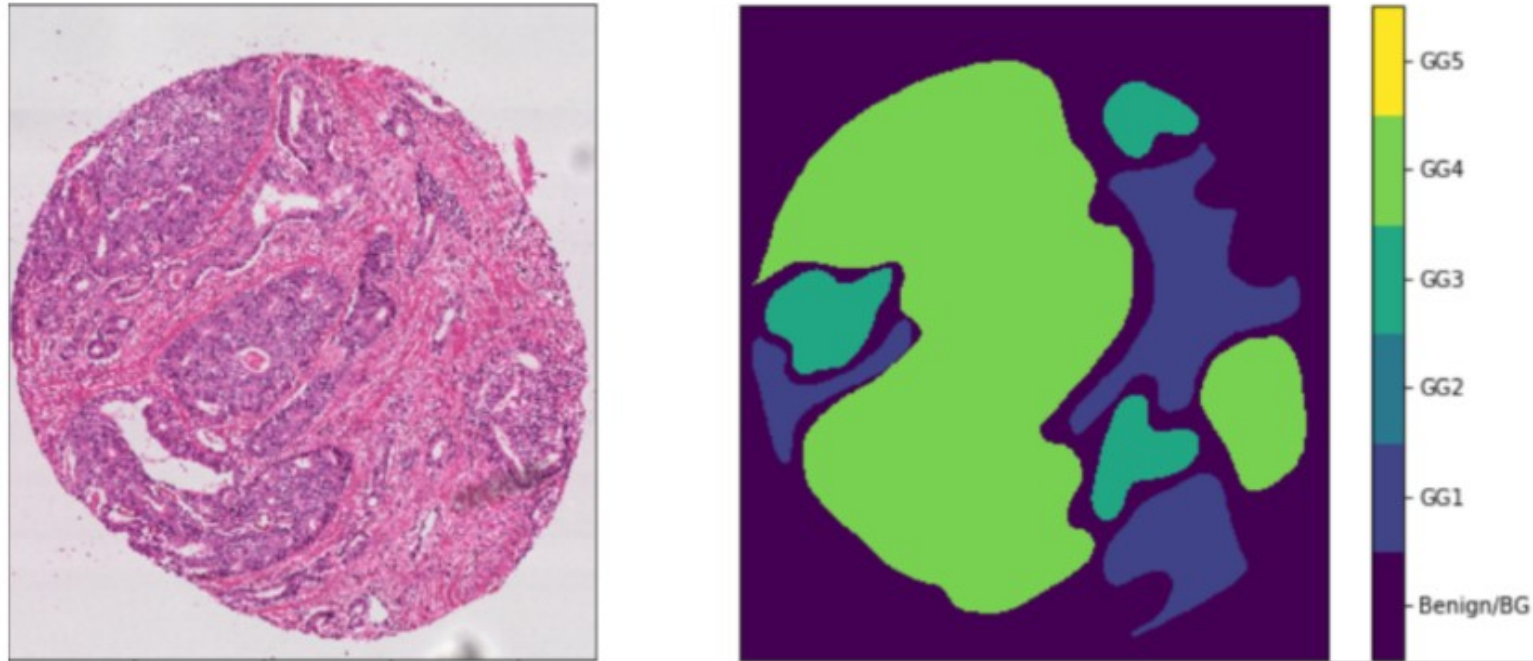


Figure 2.3: Semantic segmentation of a PCa sample, core image (left) and region segmentation and classification (right)

Human-AI interaction: an example

- But how can we be sure the labels are correct?

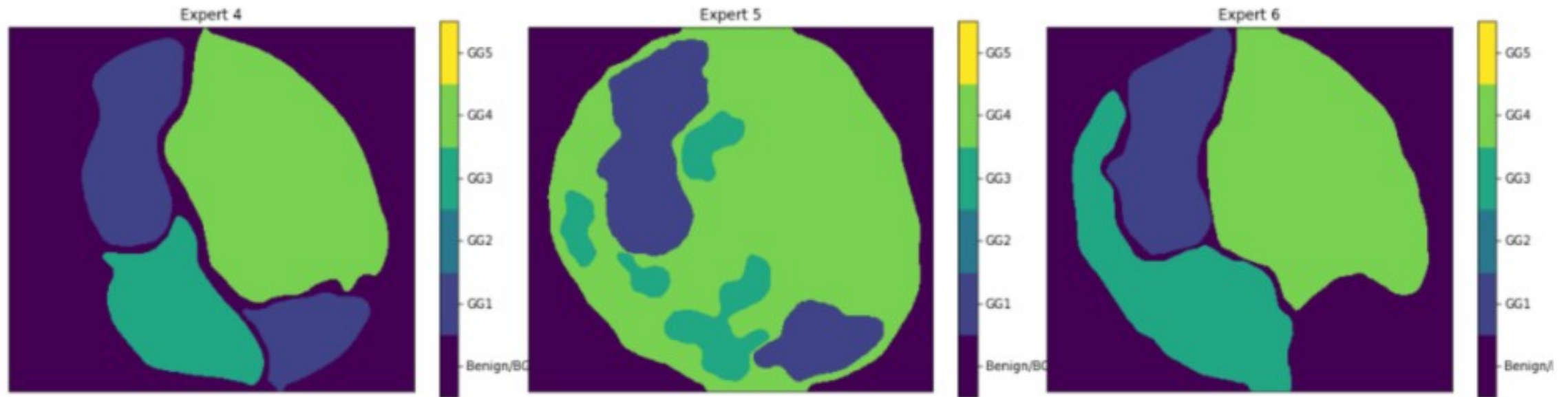


Figure 2.4: Illustration of the high-inter rater variability in PCa grading.

Human-AI interaction: an example

- The same image is labeled differently when different **experts** look at it
- More generally, real data, and particularly labels used as supervision in machine learning algorithms, can contain errors
- What kind of results can we expect from an algorithm fed with (possibly) contradictory or even wrong data?

Human-AI interaction: an example

- Possible solutions:
 - Collect correct dataset → infeasible or impossible?
 - Evaluate and correct a posteriori → infeasible or impossible?
 - **Involve human operators in the process, to control each step, to provide bidirectional feedback and explanations**
 - **Reduce the need of a single and very precise supervision and increase robustness to errors**

Human-AI interaction

- A necessary paradigm to ensure:
 - Correctness of results
 - Control of side effects
 - Adherence to goals
 - The possibility of taking action
- Hence, human-AI interaction is a key part of Trustworthy AI systems in real world applications (see also the other WPs)

Human-AI interaction

High-risk AI systems will be subject to **strict obligations** before they can be put on the market:

- **Adequate risk assessment and mitigation systems;**
- **High quality of the datasets** feeding the system to minimise risks and discriminatory outcomes;
- **Logging of activity to ensure traceability of results;**
- **Detailed documentation** providing all information necessary on the system and its purpose for authorities to assess its compliance;
- **Clear and adequate information** to the user;
- **Appropriate human oversight** measures to minimise risk;
- High level of **robustness, security** and **accuracy**.

https://ec.europa.eu/commission/presscorner/detail/en/IP_21_1682

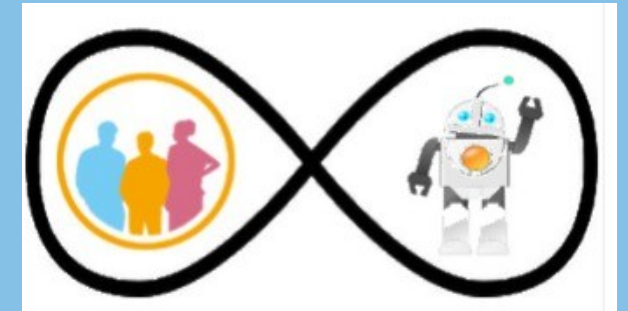
WP1: 5 tasks

1.1 Human-in-the-loop: human helps AI

- Provide continuous bidirectional feedback
- Progressively improve the datasets (labels, features, examples), their (pre)processing, and finally the AI performance

1.2 AI-in-the-loop: AI helps human

- to optimize complex systems by solving and organizing tasks in a complex workflow
- in the maintenance and adaptation of such systems in an evolving context



WP1: 5 tasks

1.3 Consensus mechanisms

- Managing conflicting opinions of human experts in an unbiased manner

1.4 Data from multiple sources and of varying quality

- Make the best use of such data, reduce the need of «perfect» annotations, increase robustness to errors

1.5 Explainable AI decision and knowledge extraction

- Increase trust in systems and their results (see also WP2)

WP1 contact

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- Suggestions?