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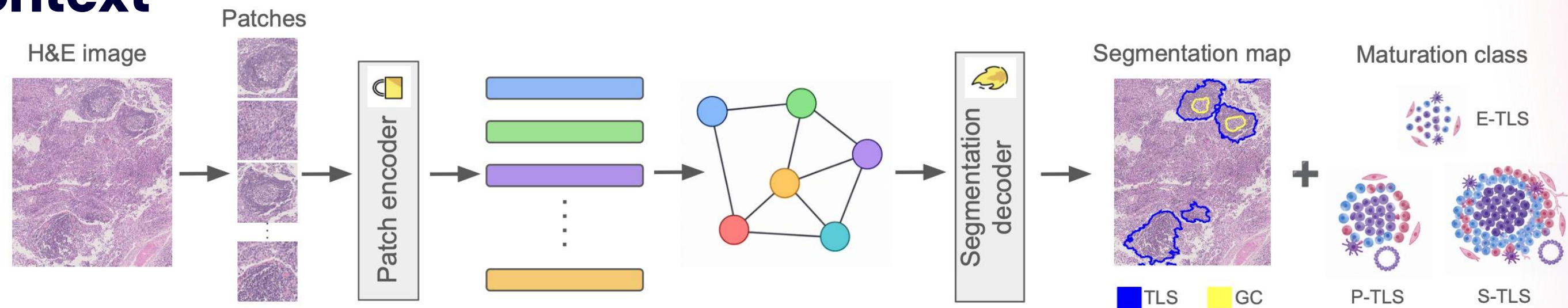


**Beyond the Patch : Context Aggregation Strategies for
Foundation Model-Driven Tertiary Lymphoid Structure
Segmentation**

Project 11

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Context



- **TLSs & germinal centers (GCs):** key biomarkers for cancer prognosis & immunotherapy response.
- **Goal:** automated TLS/GC segmentation & classification on routine H&E whole-slide images (WSIs).
- **Current baseline:** data-intensive (~1,000 annotated TCGA slides), poor cross-cohort generalization.
- **Foundation models (FM):** may cut annotation needs, but underexplored for dense segmentation.
- **Key question:** implicit (cross-attention) vs. explicit (graph-based) context aggregation.

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Founding Team

- Isinsu Katircioglu (isinsu.katircioglu@chuv.ch) – Team Leader
 - Institution: CHUV
 - Expertise: Computer vision (object segmentation & tracking), self-supervised learning, vision–language foundation models, multimodal data integration.
- Caner Ercan (caner.ercan@chuv.ch) – Team Leader
 - Institution: CHUV
 - Expertise: Computational & translational molecular pathology; spatial modeling of the tumor microenvironment to identify outcome-relevant features in pathology images.
- Additional members to be recruited at TReC'26

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Work Plan

- **WP1** — Build & benchmark FM-based context aggregation (Week 1)
 - Task 1: Build the **cross-attention strategy** on FM encoders. Compare a single-branch model against a multi-resolution dual-branch model, and test frozen encoders versus LoRA fine-tuning.
 - Task 2: Build **graph-based strategy** using patch graphs with FM node embeddings and different attention-based neighborhood aggregation.

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 - Task 1: Pick the best-performing WP1 variant and retrain it alongside the specialist baseline on progressively smaller labelled-slide fractions to quantify annotation efficiency.

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- **WP3** — Cross-cohort evaluation to quantify domain generalization (Week 2):
 - Task 1: Evaluate all models zero-shot on the in-house cohort.

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Expertise Sought

- **Required:**

- Solid Python and PyTorch experience.
- Deep-learning fundamentals: CNNs, encoder-decoder architectures, embeddings/viT tokens, and a conceptual grasp of semantic segmentation.
- Comfort reading ML papers and picking up new libraries quickly.

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- **What you'll get to explore:**
 - Transformer-based visual segmentation and graph neural networks.
 - Digital pathology basics: whole-slide image handling with OpenSlide/Lazyslide/AnnData.
 - Model benchmarking and evaluation.

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