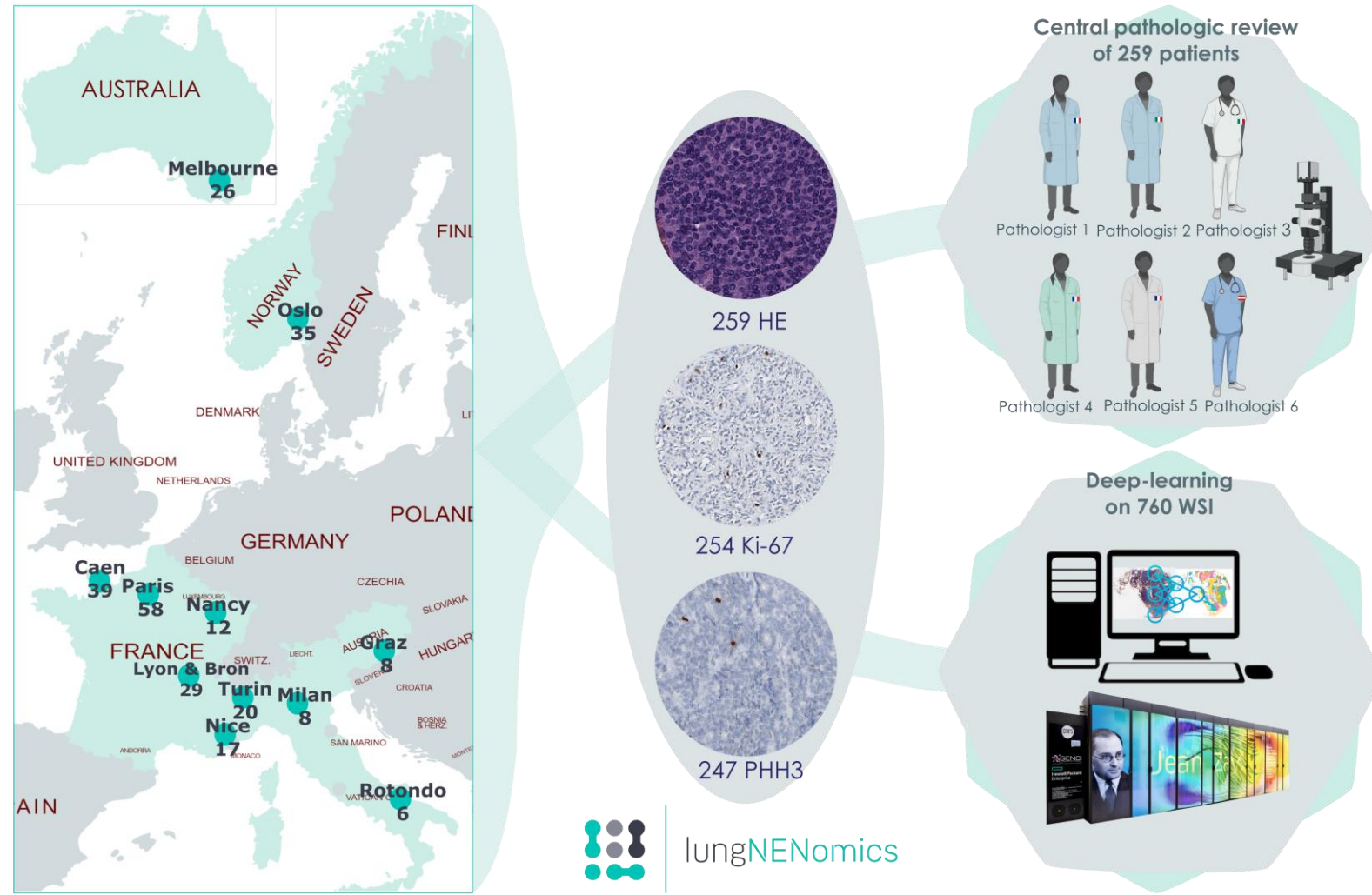


**Background:**



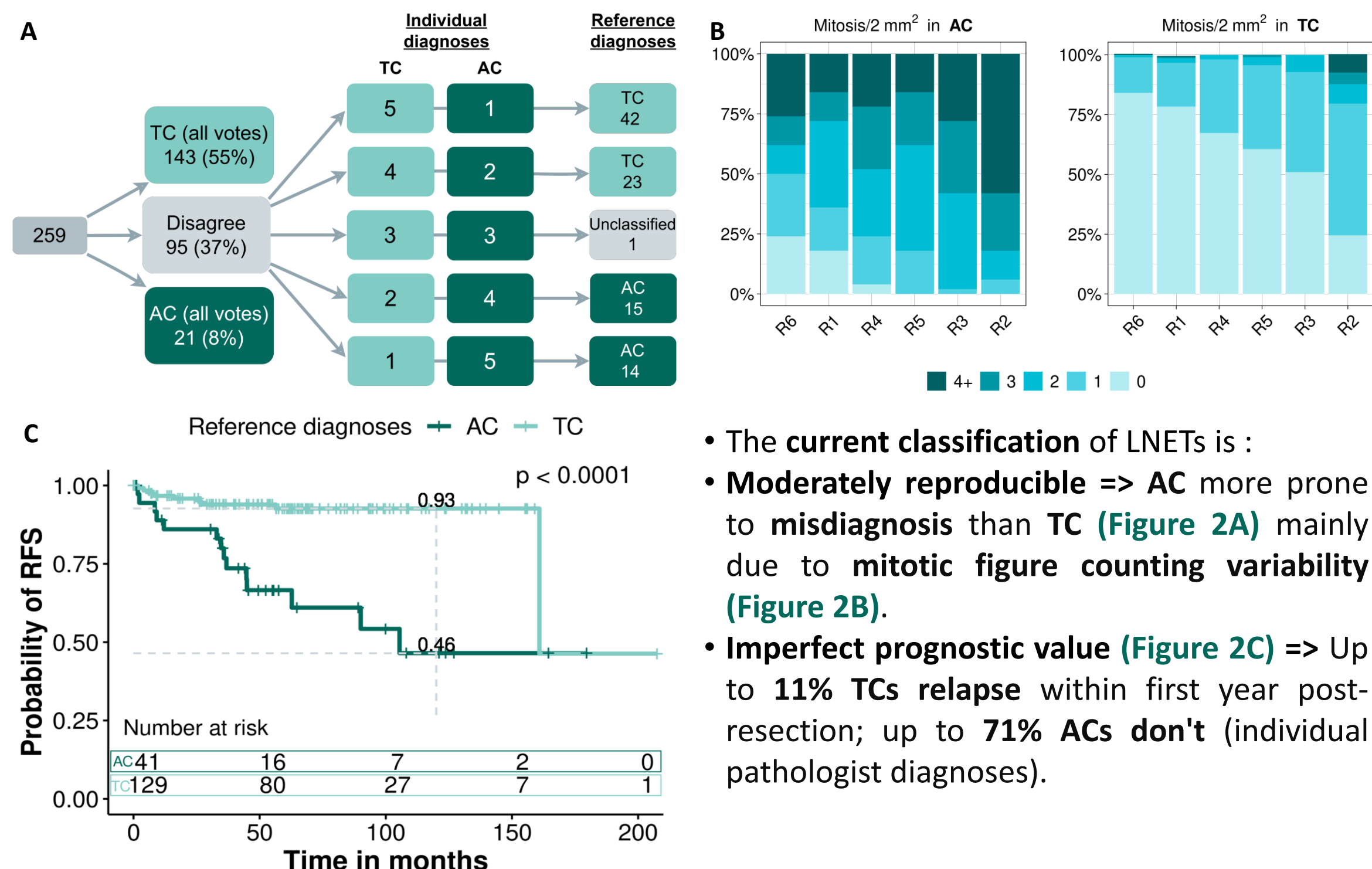
**Figure 1:** Six thoracic pathologists diagnosed 259 LNETs per 2021 WHO criteria, distinguishing ACs from TCs by focal necrosis and/or 2-10 mitoses per 2 mm<sup>2</sup>.

The lungNENomics cohort presents a unique opportunity to assess the strengths and limitations of current WHO classification and to evaluate the utility of emerging markers

**Methods:**

- Tumour proliferative activity: **Ki-67 index** and **phospho-histone H3 (PHH3)** protein expression => both quantified by each pathologist in hot-spot areas + automatically by the Pathonet [Negahbani SCI REP 2021] deep learning models at the whole slide image (WSI) scale.
- **Hypothetical classification systems** generated for each marker and nine cut-off values, replacing mitotic count. Prognostic values of resulting groups were assessed based on RFS.
- The **unsupervised Barlow twins** [Zbontar PMLR 2021, Quiros Arxiv 2022] deep learning model was used to identify potential new morphological features.

**Result 1: Limitations of the current morphological criteria**



**Figure 2:** A) Flow diagram of the pathological review leading to reference diagnoses. B) Distributions of the number of mitoses by pathologist. C) Kaplan-Meier curves of recurrence-free survival (RFS).

In addition to 2021 WHO criteria distinguishing AC from TC, **we propose adding Ki67 and PHH3 IHC:**

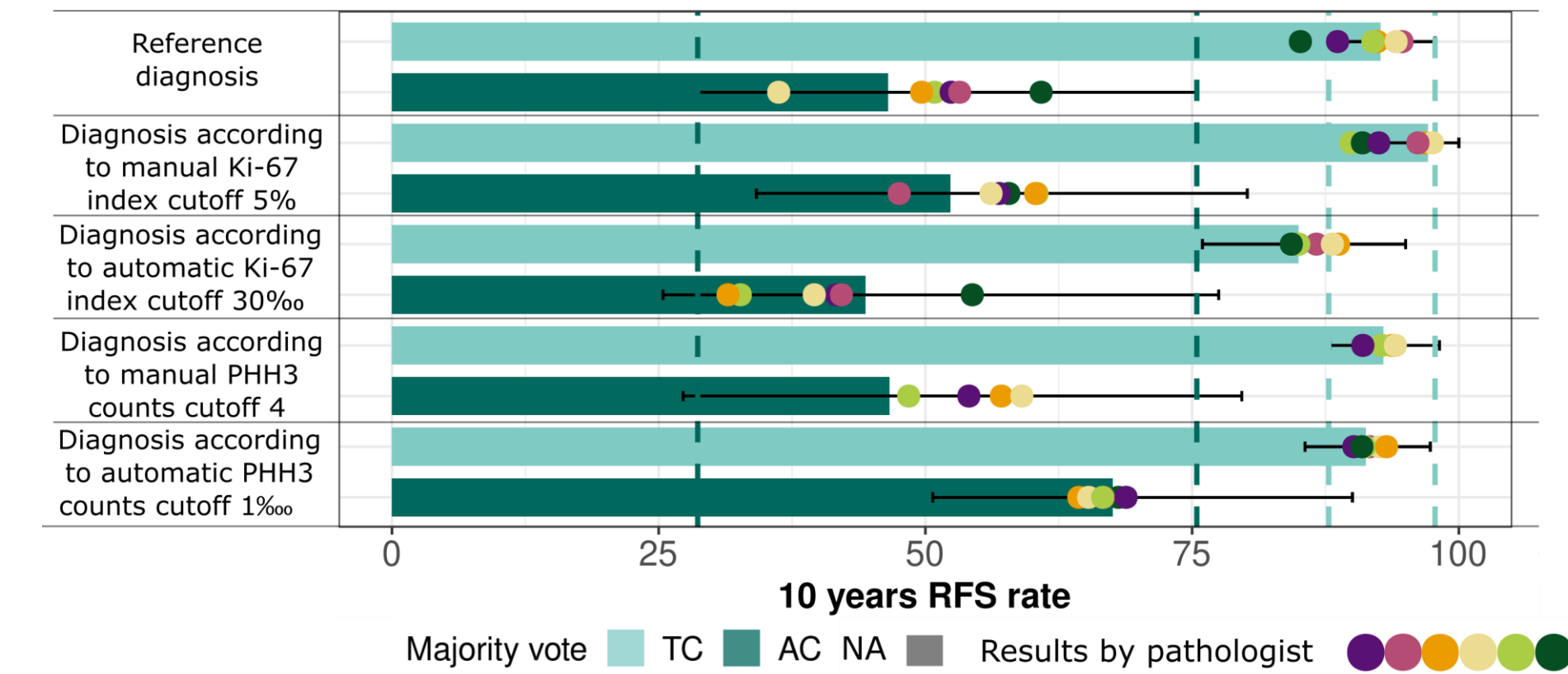
- **To guide mitotic count**, considering tumor heterogeneity and high proliferation areas;
- **To identify potential NET-G3** with carcinoid morphology (>10 mitoses/2mm<sup>2</sup> & Ki-67 > 20%);
- **To ensure consistency of mitotic count with PHH3 staining** (~2x more PHH3 positive-cells expected than the number of mitotic figures on HE);
- **To detect possible AC when Ki67 reaches 5%** either on biopsy or when the mitotic count is not feasible;

Perspectives: Multi-omics data analysis reveals robust molecular groups with varying morphological features using unsupervised deep-learning methods, paving the way for morpho-molecular classification.

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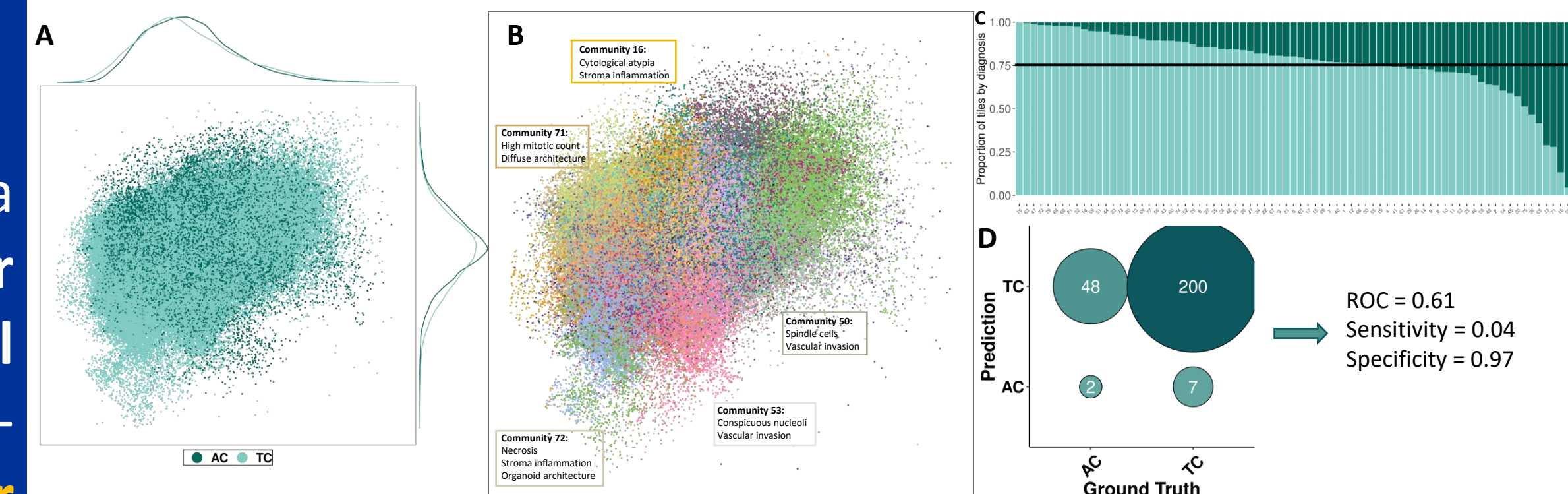
**Result 2: Added value of Ki-67 and PHH3 in the assessment of proliferative activity**



**Figure 3:** (Left) For each classification system, the coloured bar corresponds to the 10-year RFS rate based on the diagnoses resulting from majority voting.

- A wide range of thresholds applied to pathologists' estimates of Ki-67 and PHH3 expression enable LNETs to be divided into two groups with similar prognostic values to those defined by the WHO criteria (Figure 3).
- Ki-67 improves high-risk TC detection but not AC specificity.
- PHH3 speeds up mitotic counting but does not significantly reduce inter-observer variability.
- Automatic evaluation of Ki-67 and PHH3 reach the performance of expert pathologists (Figure 3).

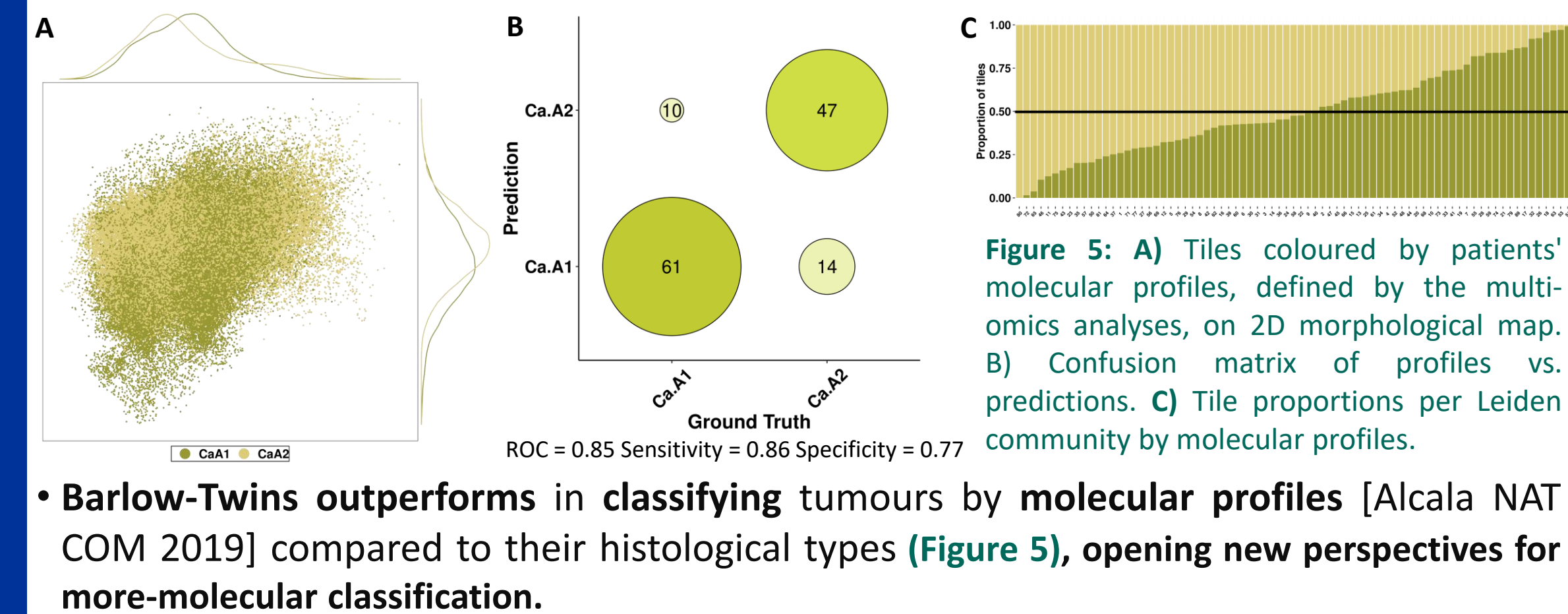
**Result 3: Deep learning to uncover novel clinically relevant morphological features**



**Figure 4:** Unsupervised deep-learning experiment on HE/HES WSI. A) 2D morphological map color-coded by patient diagnosis. B) Leiden communities annotated for morphological enrichment. C) Tile proportions per Leiden community by tumor type. D) Confusion matrix & performance metrics.

- State-of-the-art deep learning on WSI doesn't aid TC vs. AC distinction due to histological similarity (Figure 4), hinting at morphology's limitations.

**Future Directions for Research:**



- Barlow-Twins outperforms in classifying tumours by molecular profiles [Alcalá NAT COM 2019] compared to their histological types (Figure 5), opening new perspectives for more-molecular classification.