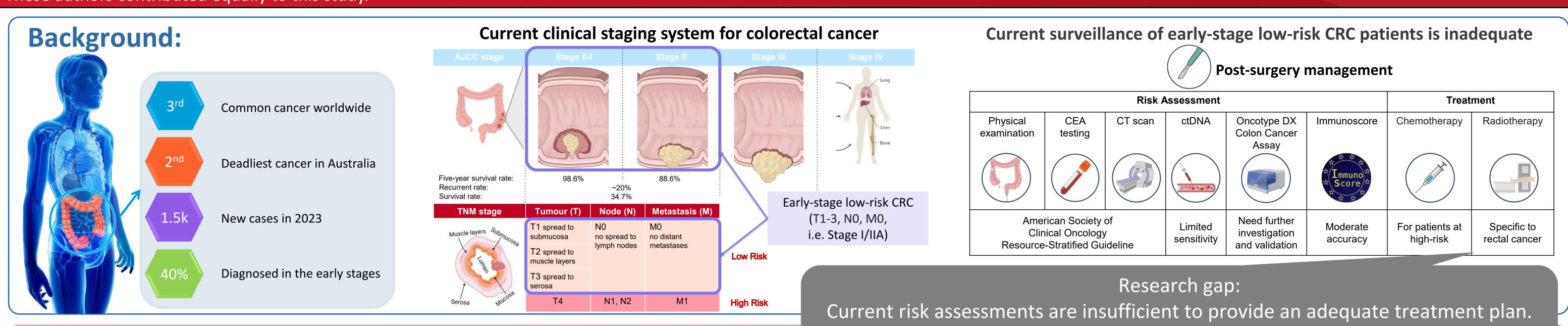
# Developing Al approaches to unravel factors that drive disease relapse in low-risk, early-stage colorectal cancer

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# **Hypothesis and Objective:**

Develop a custom model for

tumour type based on whole

Develop custom models for

the time to relapse of CRC

automated detection of

slide images (WSIs).

Aims:

Aim 1

Aim 2

patients.

Aim 3

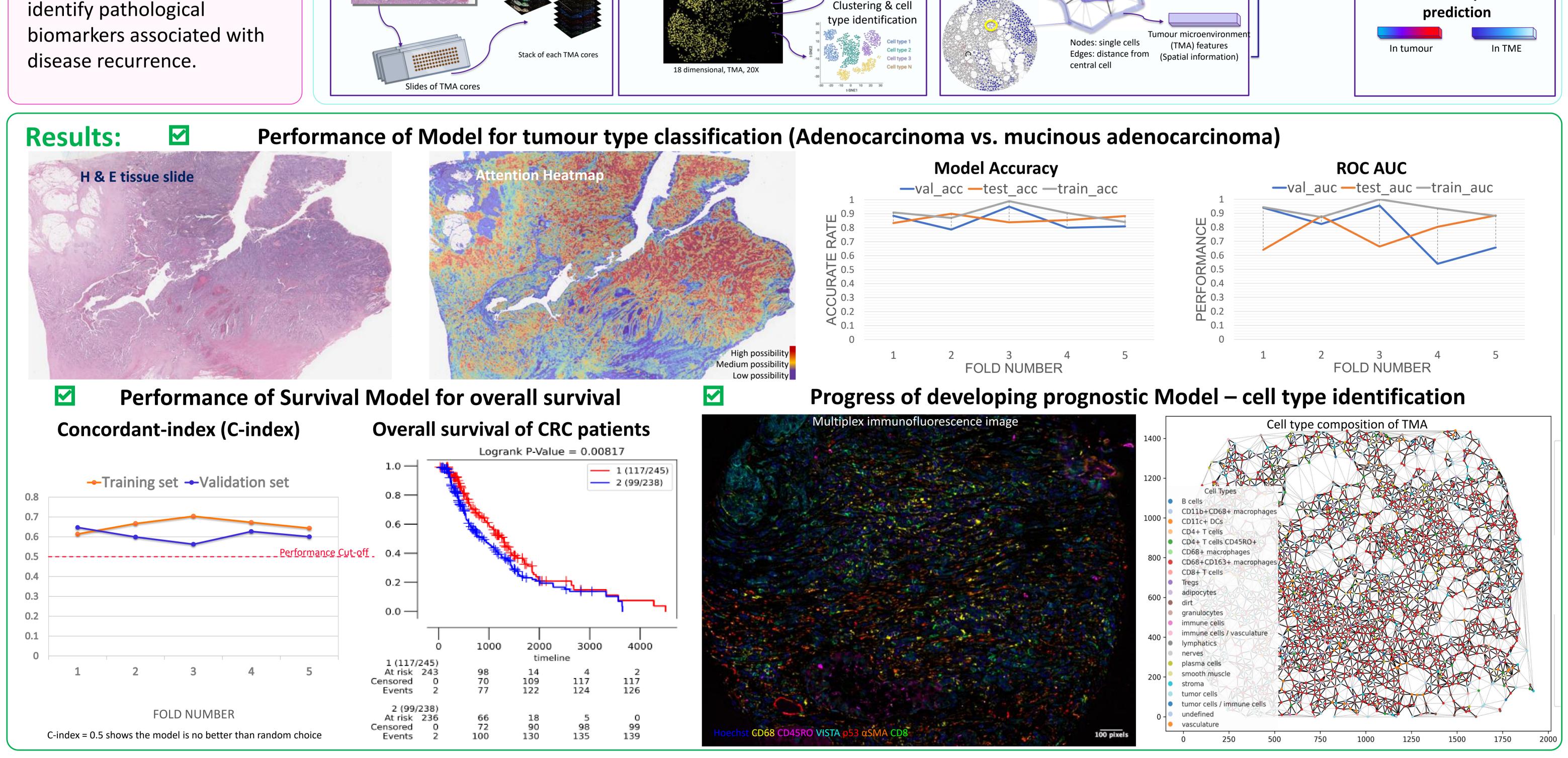
Hypothesis: Tumour tissues with biomarker signatures that drive recurrence predispose some patients to disease relapse.

Objective: To develop a multi-modal data-driven AI approach for CRC prognostic prediction and identification of biomarkers associated with CRC relapse.

#### Pipeline for whole slide image pre-processing Model developing for WSIs **Outputs Network construction** WSI acquisition Image pre-processing H&E slide Feature extraction **Clinical prognostic** Graph convolution network factors (GCN) construction Central features ដំហំជា 🕹 🖔 🕻 🖆 Local information) Segmentation Surrounding features (Spatial information) **Deep-learning** Patch features **Model training** Deep-learning network **Prognosis prediction** Pipeline of multiplex immunofluorescence images Model developing for mIFs predicting overall survival and Graph neural network (GNN) construction Image acquisition Image pre-processing **Expression of biomarkers** Registration 3-hop subgraph Recurrence interva Integrate generated models to Voronoi graph (Local information) Cores of TMA **Biomarker expression** Clustering & cell prediction type identification Stack of each TMA cores Edges: distance from (Spatial information)

This is partly due to our limited knowledge of underlaying mechanisms and the

associated biomarkers that contribute to disease recurrence.



### **Conclusion and Future Directions:**

- We developed a deep-learning network to classify CRC tumour type, as well as a deep-learning network to predict the overall survival of CRC patients using whole slide hematoxylin and eosin (H&E) images. The Graph neural network model for prognostic prediction using multiplex immunofluorescence images has made progress in identifying cell types.
- Future integration of these models will allow for the analysis of the saliencies of prediction outcomes, as well as the detection of biomarkers associated with the underlying

### Contact



## References

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