

Clinical Background [1][2][3]:

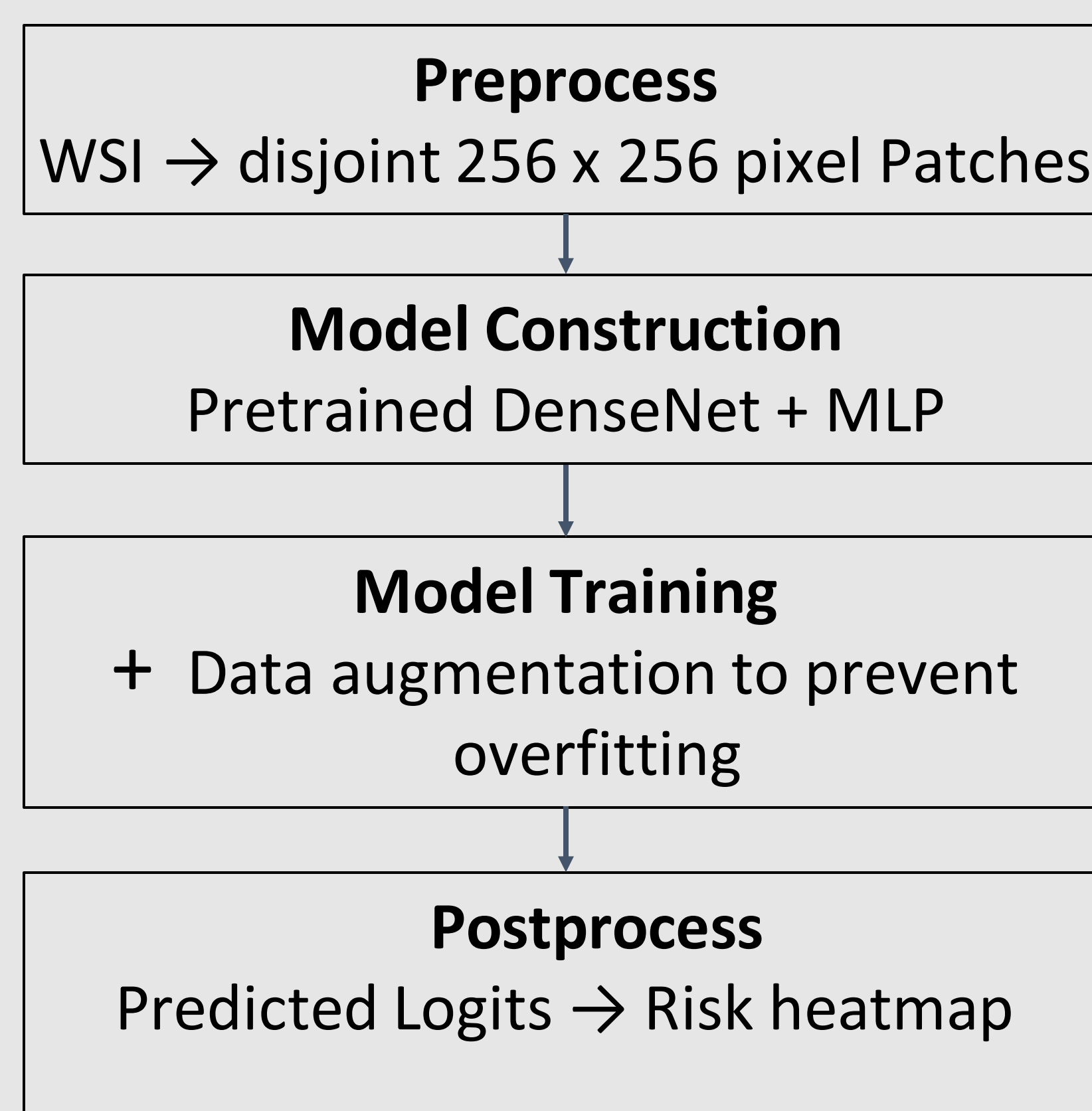
- Prostate cancer is the #2 killer of men worldwide
- Gleason Grading is a reliable method of determining the severity of prostate cancer and planning treatment
- Estimation of Gleason grades requires expert pathologists → problem for low-resource settings

Problem Statement:

Deep learning image analysis can be a useful tool, but previous models are

- Non-generalizable
- Trained on small datasets
- Don't account for differences in pathologist determination.

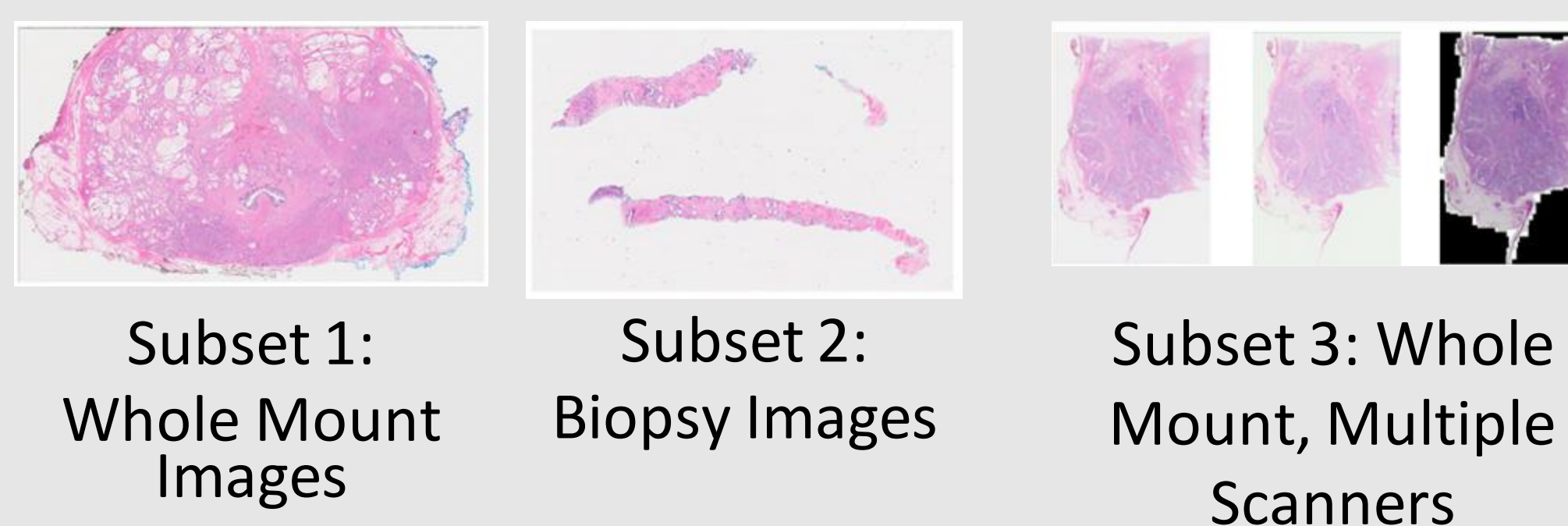
Method:



Dataset:

AGGC 2022 dataset [4]

Ground truth provided by multiple pathologists



Conclusions:

We developed a deep learning model to automatically identify all five Gleason patterns with an accuracy of 0.74

Future Direction:

- Determine generalizability of methods/model to other cancer classification tasks
- Determine on clinical utility of final product, refine model and outputs accordingly

DEEP LEARNING METHODS FOR AUTOMATED GLEASON GRADING

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Results:

① Subset-dependent model performs better than Subset-independent model

	Training Resources	Accuracy	Averaged
Subset-dependent	Subset1	0.67	0.74
	Subset2	0.79	
	Subset3	0.75	
Subset-independent	All Subset	0.66	0.66

② Morphological transformation further improved results

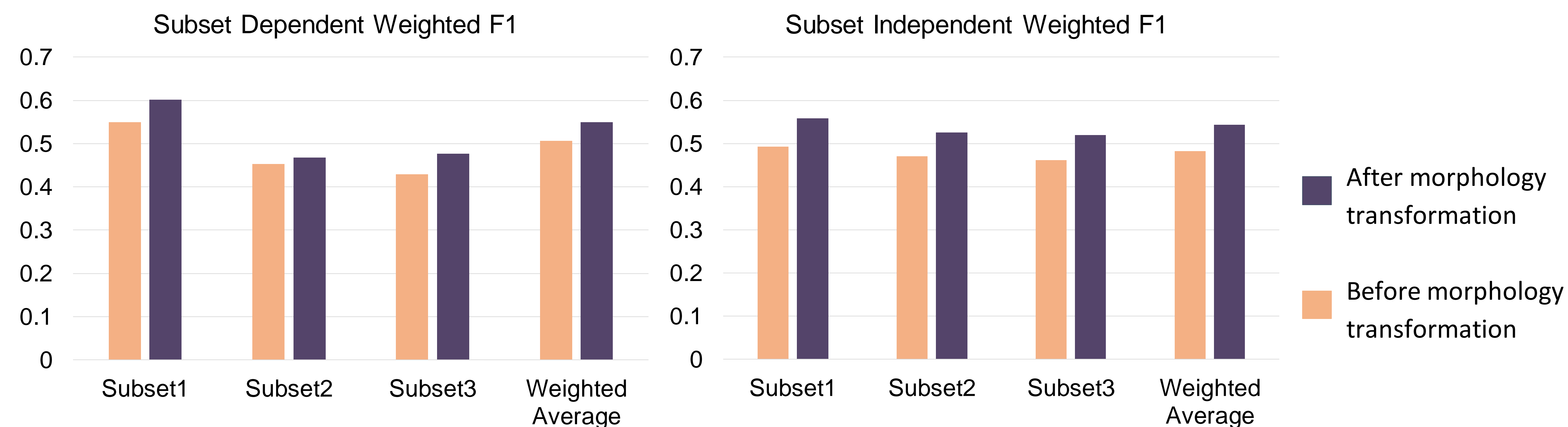


Figure 1: Comparison of weighted F1 with and without morphology transformation.

$F1 = 2 \times \text{Precision} \times \text{Recall} / (\text{Precision} + \text{Recall})$; $\text{Precision} = \text{TP} / (\text{TP} + \text{FP})$; $\text{Recall} = \text{TP} / (\text{TP} + \text{FN})$

Subset-wise Weighted F1 = $0.25 * F1_{G3} + 0.25 * F1_{G4} + 0.25 * F1_{G5} + 0.125 * F1_{Normal} + 0.125 * F1_{Stroma}$

Total Weighted Average F1 = $0.6 * \text{weighted F1}_{subset_1} + 0.2 * \text{weighted F1}_{subset_2} + 0.2 * \text{weighted F1}_{subset_3}$

③ Predicted heatmap shows good alignment with ground truth

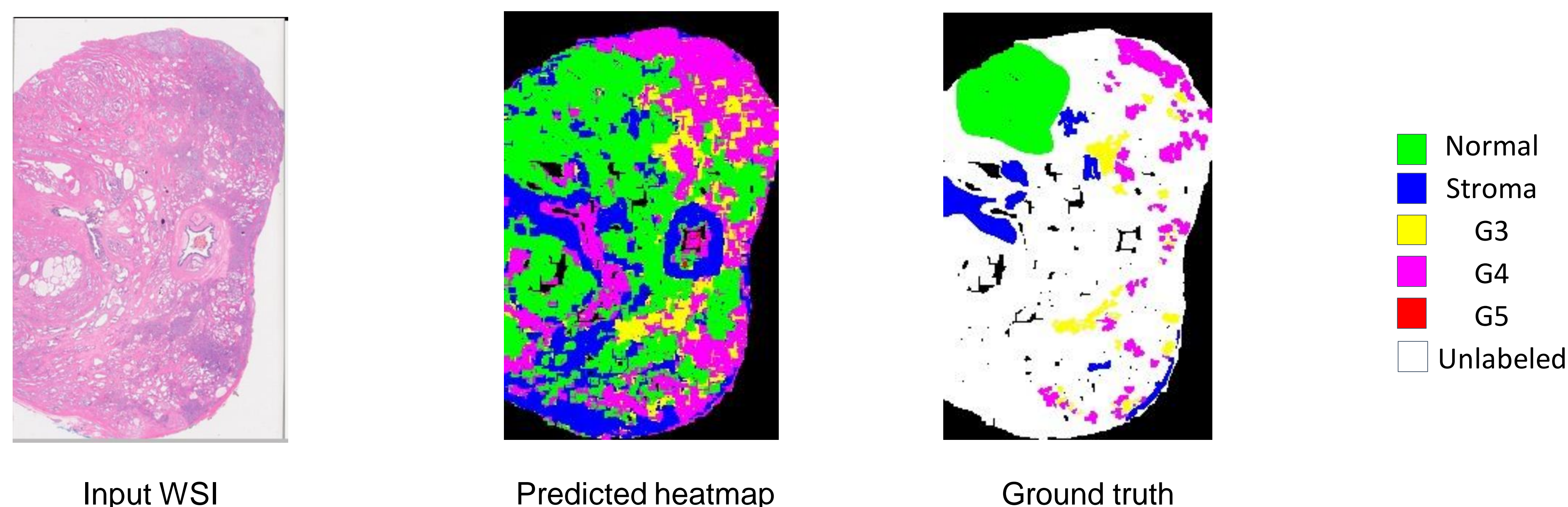


Figure 2: An example of the predicted heatmap in comparison with ground truth. Heatmap opaqueness was adjusted by confidence level, where more transparent regions indicate lower confidence level.