

## Problem Definition and Contribution

**Goal:** Generating multi-stained histological images utilizing correlations across adjacent slices on the basis of existing unpaired virtual staining methods.

### Key Contributions:

- A novel **adjacency-guided multi-domain transfer framework (AGMDT)** to transfer renal histology images into multiple staining types.
- **Adaptive Pairing Module** and **Adjacency-guided Encoder** which can avoid pixel-level alignment and meanwhile utilize the correlations among adjacent tissue slices.
- A **renal histological dataset** where each case possesses all the four staining types: H&E, Masson, PASM, PAS that applied on serial slices of the same tissue.

## Dataset

After the processing of the adaptive pairing module, we have successfully created a comprehensive renal histological dataset.

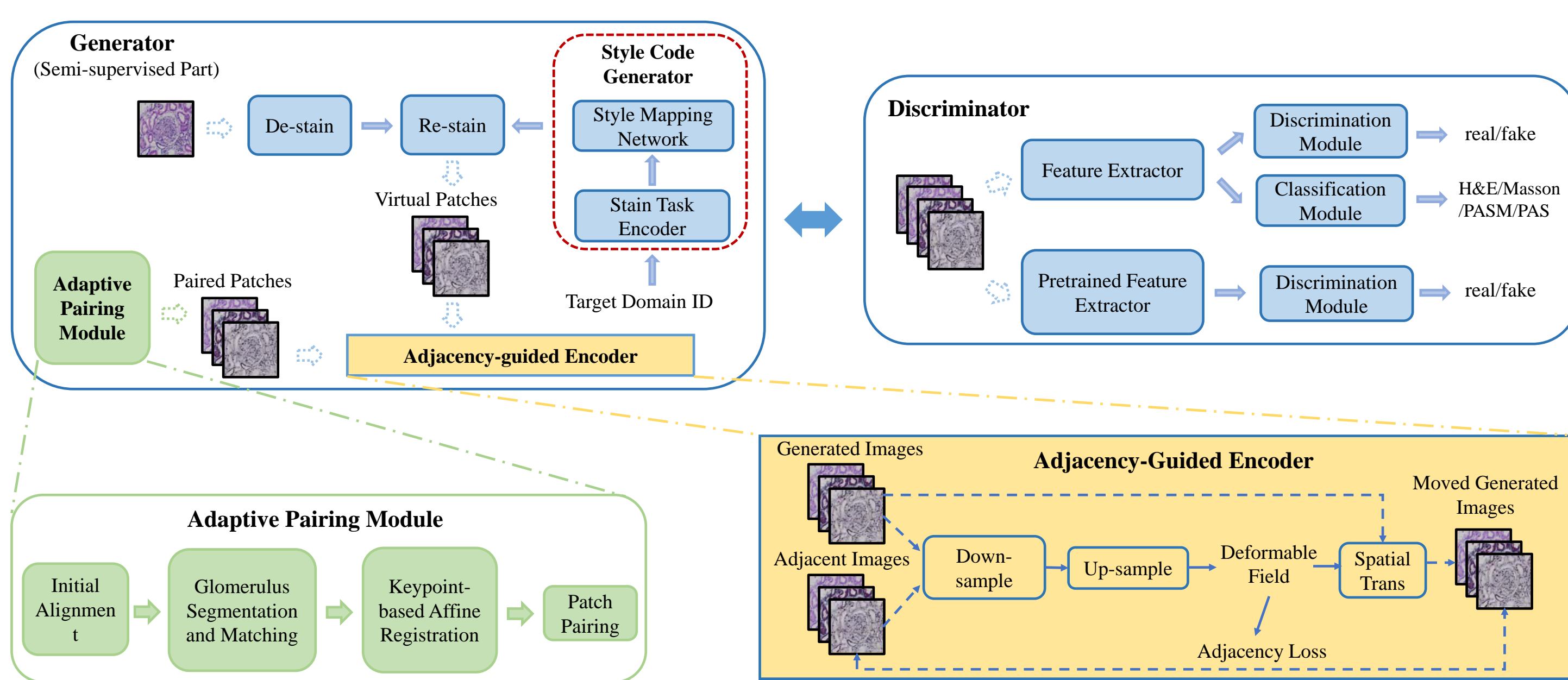
### Details

- Obtained from serial slices stained with H&E, PAS, PASM, and Masson.
- 188 whole slide images from 22 patients with pathologists' WSI-level diagnoses.
- 32,413 pairs of patch-level aligned glomeruli.

## Method

### Overall Structure of AGMDT:

Adjacency-Guided Multi-Domain Transfer Framework



- AGMDT has two key components, respectively an adaptive pairing module and an adjacency-guided encoder.

### Loss function:

$$\begin{aligned} \mathcal{L}^G = & \lambda_1 \times (\mathcal{L}_{adv}^G + \mathcal{L}_{cls}^G + \mathcal{L}_{cyc} + \mathcal{L}_{idt}) \\ & + \lambda_2 \times (\mathcal{L}_{\eta adv}^G + \mathcal{L}_{\eta cls}^G) + \lambda_3 \times \mathcal{L}_{adj} \end{aligned} \quad (1)$$

The generator's loss function consists of six components, including:

- adversarial loss  $\mathcal{L}_{adv}^G$
- classification  $\mathcal{L}_{cls}^G$
- auxiliary loss  $\mathcal{L}_{\eta adv}^G, \mathcal{L}_{\eta cls}^G$
- cycle loss  $\mathcal{L}_{cyc}$
- identity loss  $\mathcal{L}_{idt}$
- adjacency-guided loss  $\mathcal{L}_{adj}$ , which measures the discrepancy between the generated data and its corresponding adjacency data.

## Experiments & Results

### Blind Evaluation:

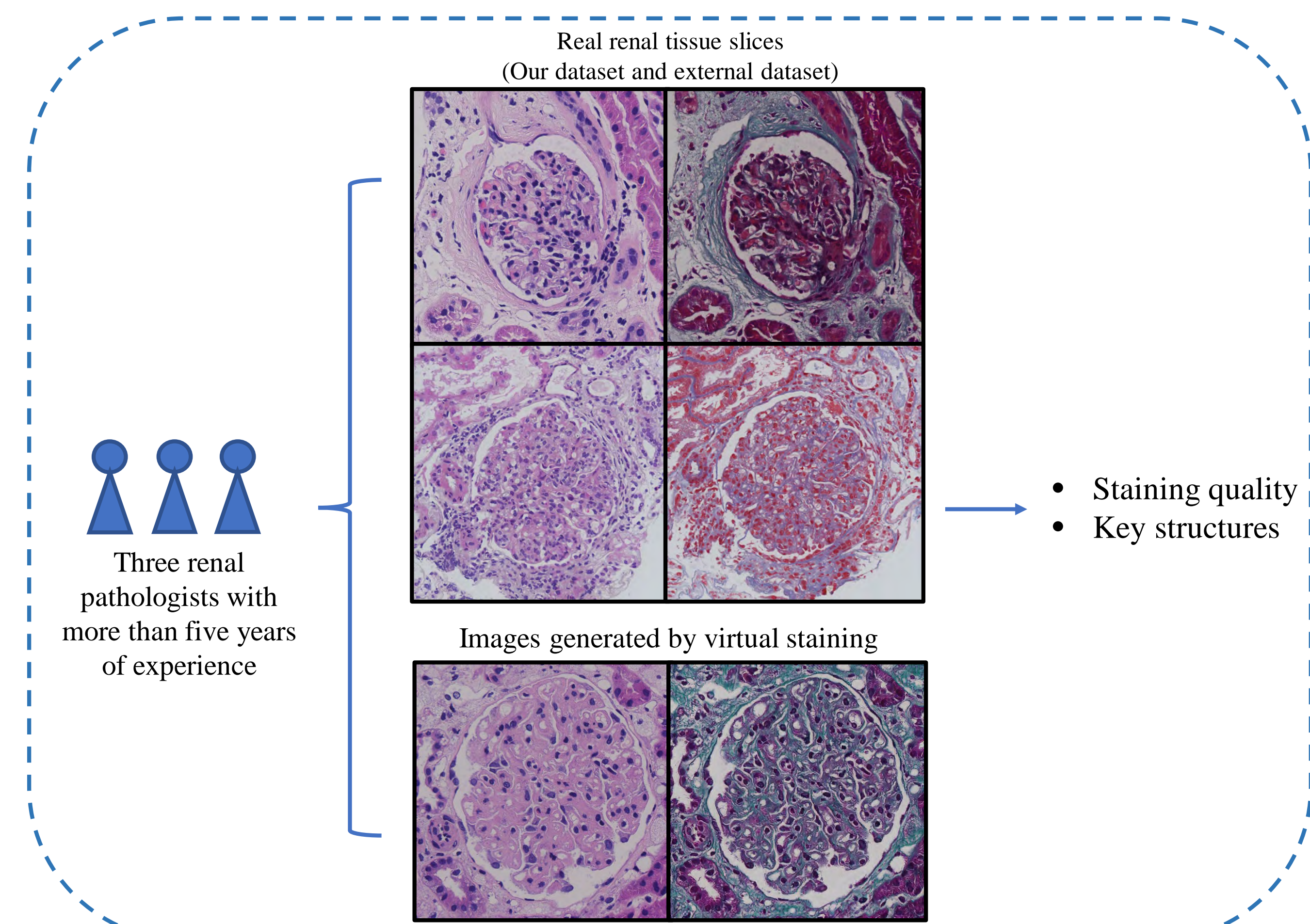


Table 1: Comparison of different methods

Methods	H&E2MASS		H&E2PASM		H&E2PAS	
	DISTS↓	DBCNN↑	DISTS↓	DBCNN↑	DISTS↓	DBCNN↑
FUNIT[4]	0.2452	56.52	0.2496	54.01	0.2259	53.17
MUNIT[1]	0.3207	29.76	0.3037	32.67	0.2764	36.02
UGATIT[2]	0.2374	57.18	0.2467	55.64	0.2166	51.91
UMDST[3]	0.2064	53.44	0.2770	54.02	0.2162	51.90
Ours(w/o Adj)	0.2143	58.75	0.2468	60.61	0.1917	55.74
Ours (w Adj)	<b>0.2020</b>	<b>64.49</b>	<b>0.2453</b>	<b>66.23</b>	<b>0.1909</b>	<b>59.85</b>

Table 2: Blind Evaluation of staining results

	Masson	PASM	PAS
Real(our part)	4.25	<b>4.92</b>	4.33
Real(external part)	3.48	4.77	<b>4.46</b>
FUNIT[4]	0.69	1.42	2.25
MUNIT[1]	2.67	0.75	2.42
UGATIT[2]	2.67	3.67	3.08
UMDST[3]	4.10	4.09	4.17
AGMDT	<b>4.59</b>	4.17	4.25

The virtual staining results of AGMDT and baseline methods:

### Ablation of Adjacency-guided Encoder:

