

# PROBLEM

## INEFFICIENT CANCER DETECTION METHODS

## COSTLY DIGITAL PATHOLOGY SCANNERS AND PATHOLOGIST DIAGNOSTICS

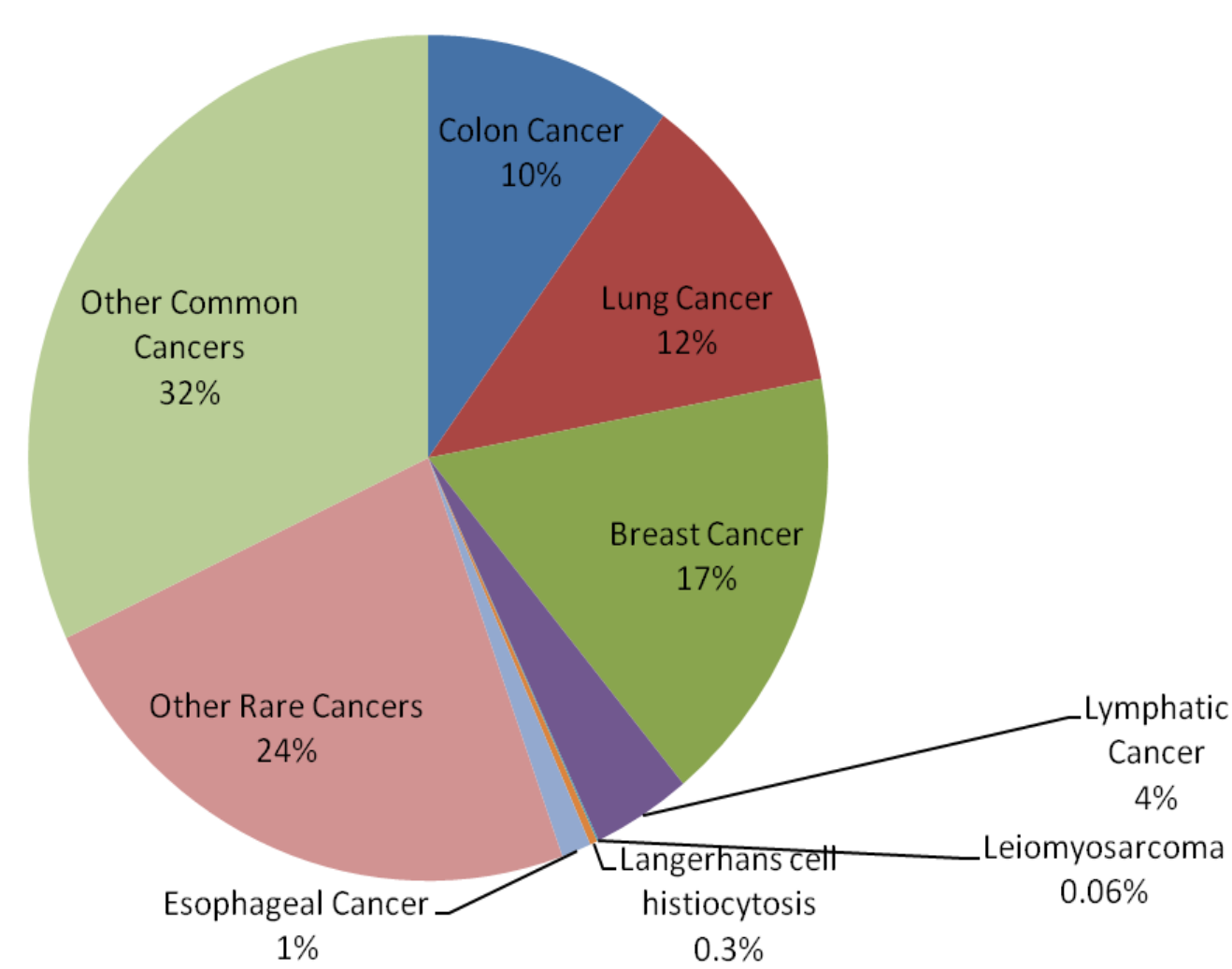
## CANCER DIAGNOSTIC DISPARITIES

## LOW-INCOME AREAS CANNOT AFFORD QUALITY CANCER DIAGNOSIS.

- Rare cancers and cancer adjacent diseases such as leiomyosarcoma, esophageal cancer, and Langerhans-cell histiocytosis **make up 25% of all cancers in the U.S.** (NIH)
- However, individually, rare cancers are defined as cancers that affect less than 40,000 people per year
- This makes it incredibly difficult for doctors to diagnose such cancers due to the lack of information and data needed to identify signs of a tumor
- This issue is **exacerbated further in low-income and rural areas** due to being exorbitantly expensive for some individuals, requiring highly trained doctors, and requiring costly equipment
- The **gold** standard for tumor diagnosis is tissue biopsy and analysis by a pathologist with a histopathology scanner
- However, the equipment and finances needed to fund a single scanner **can reach up to \$300,000** (Digital Pathology Association), making it incredibly difficult for developing areas to purchase and receive high-quality cancer diagnoses
- Patients may also have to individually **pay up to \$3000 for a full procedure** (Turquoise Health)

A cheaper and more accurate rare cancer detection system must be built

Selected Cancers and Associated Prevalance



Makeup of all cancers as compared to selected cancers. Image Credit: Uddip Kashyap

# ENGINEERING GOALS

- Cost**
  - The device should be of a cost less than or equal to \$500
- Quality**
  - The device should be able to capture high-resolution images at different magnifications
- Viewing**
  - The device should have software that enables images to be stitched together for whole-slide image (WSI) viewing
- Usability**
  - A user-friendly application must be developed, and the software must be compatible with said application

# PathScan: A Digital Pathology Pipeline for Efficient Cancer Diagnostics

Uddip Kashyap

## METHODOLOGY: SCANNER

### 1 ProGAN Generation

- Upscale images from 4x4 pixels to 32x32 pixels
- Generate synthetic datasets of 100 images using **random noise vectors**

### 2 CNN and Fine-tuning

- Tune ProGAN hyperparameters to **achieve FID  $\leq 250$**
- Create CNN for detection of each rare cancer type

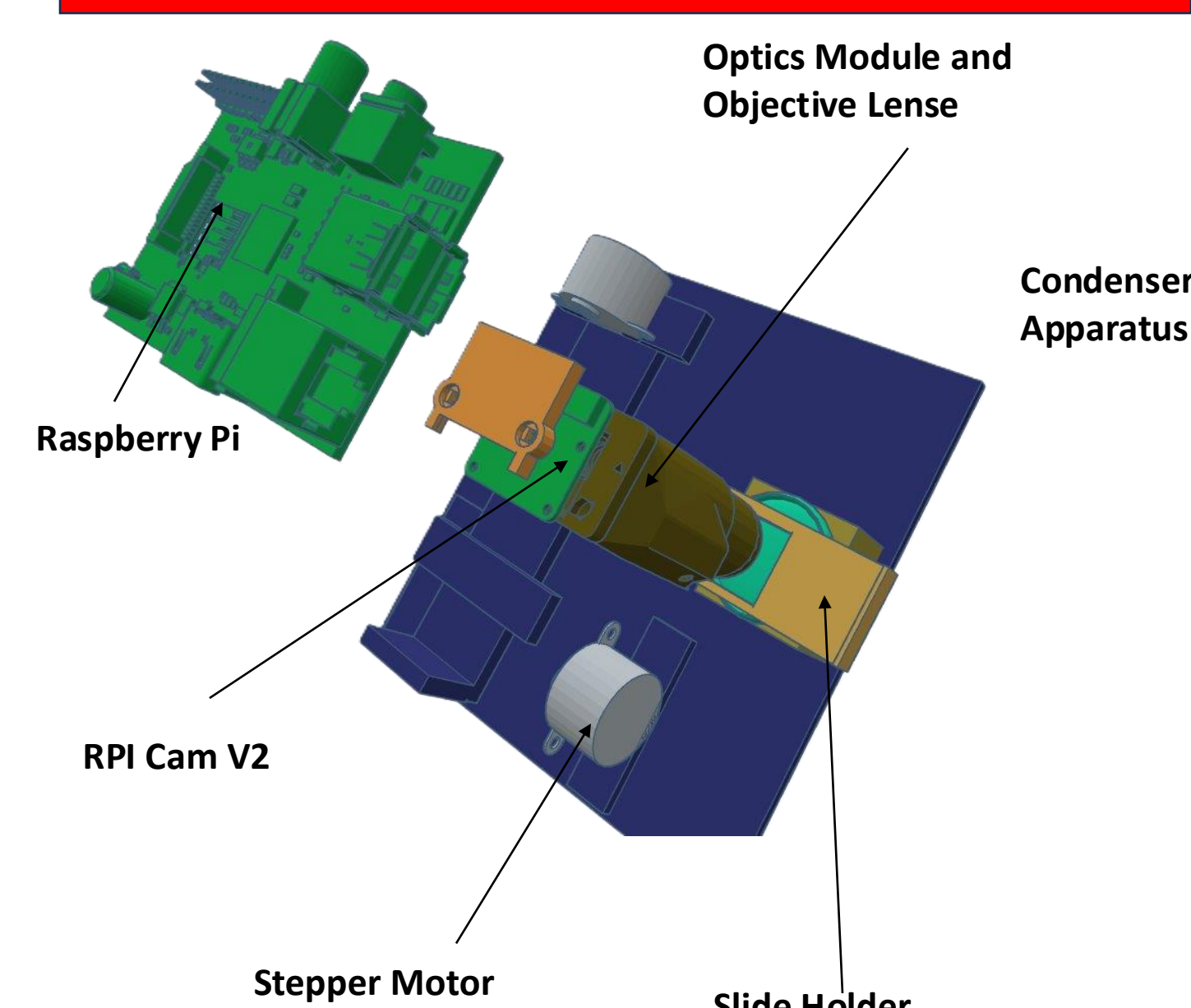
### 3 Scanner Design

- Design and develop scanner with **low-cost components**
- Capture and stich images for analysis

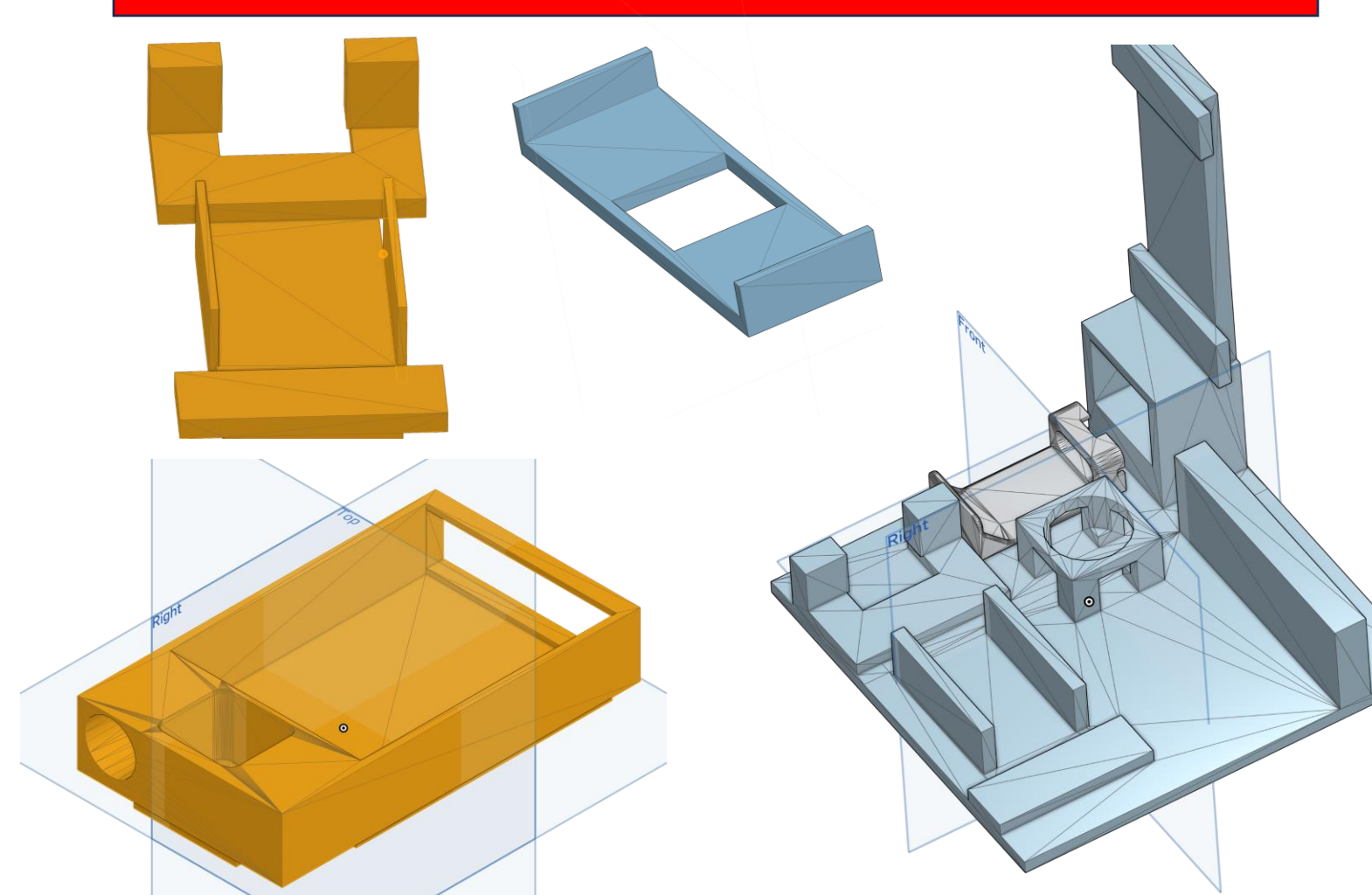
### 4 Deployment and GUI

- Simple GUI with **AI integration** for cancer detection
- Data security mechanisms to protect patient information

### PRELIMINARY CAD RENDERING

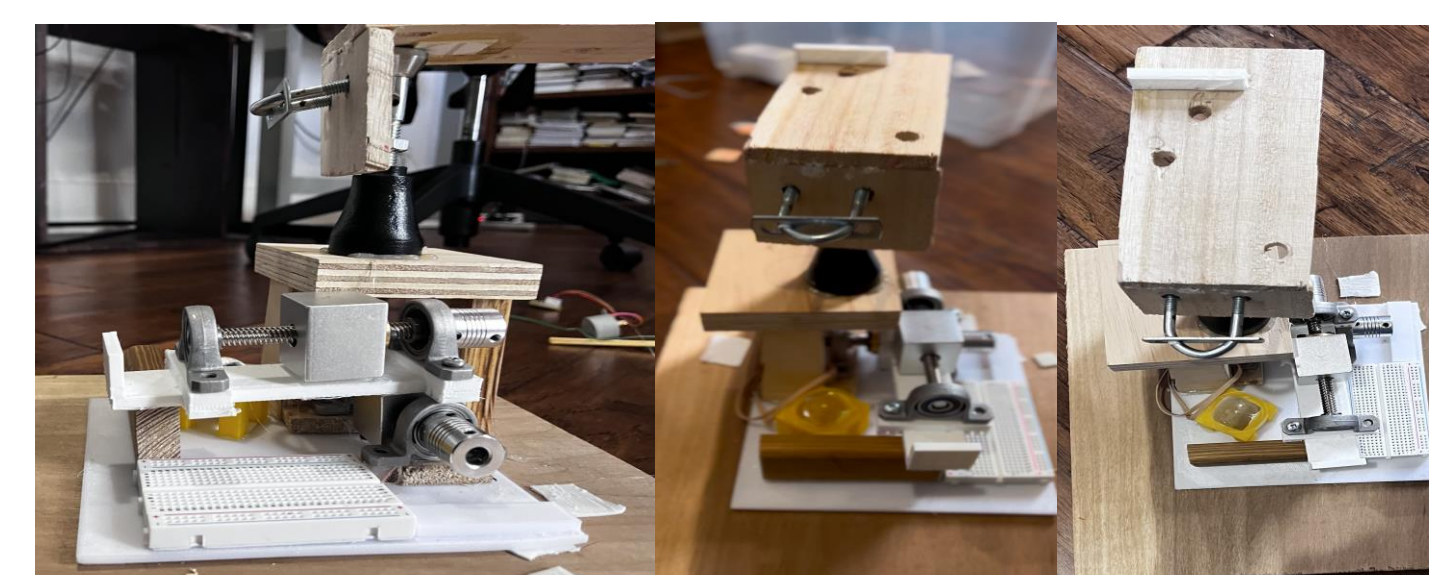


### FINAL CAD MODELS



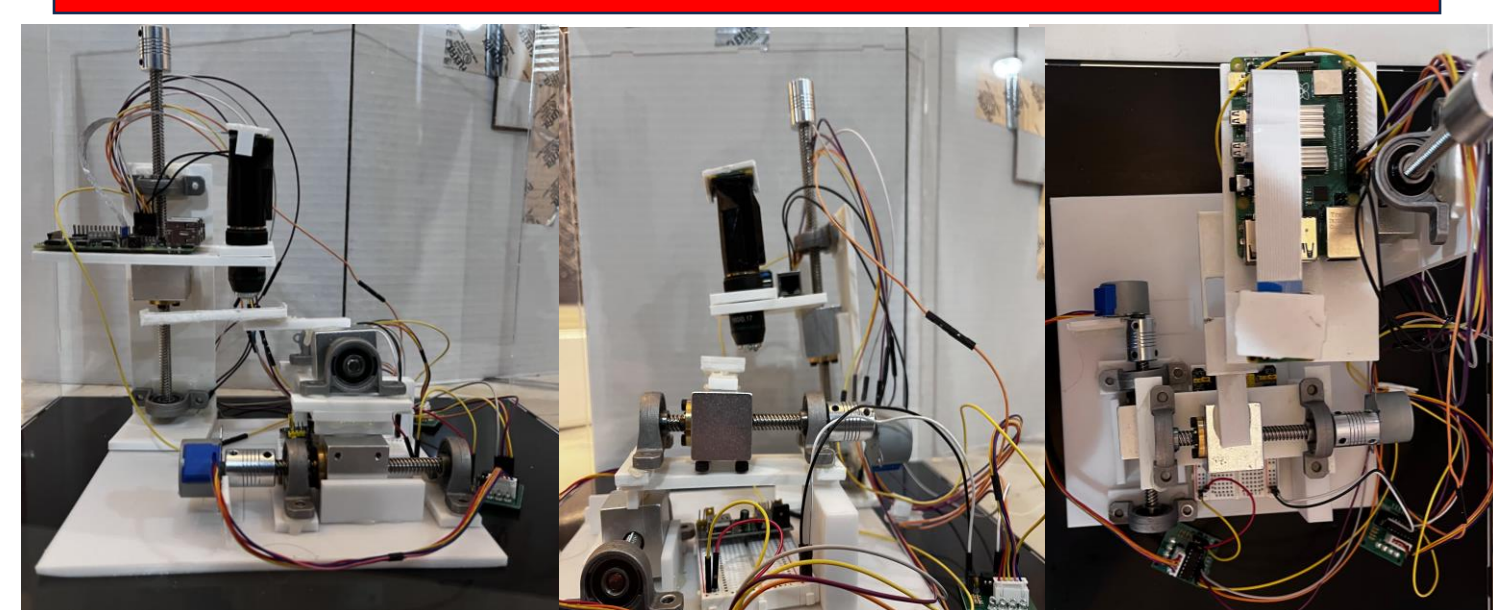
- A Raspberry Pi Camera Module is connected to a modified OpenFlexure optics module and an objective lens to **magnify and focus the slide for image capturing**
- The slide itself rests on a condenser apparatus which **focuses the beams of light onto the tissue**. The stepper motors move the slide to capture all aspects of the tissue sample
- Through a ribbon cable, images are transferred to the Raspberry Pi
- Portions of the image can be used by the PathScan Histology Classifier for an **independent diagnosis**
- Over **20 trials**, the scanner averaged **10 minutes** per scan, including computation times, showing high efficiency

### PRIMARY PROTOTYPE



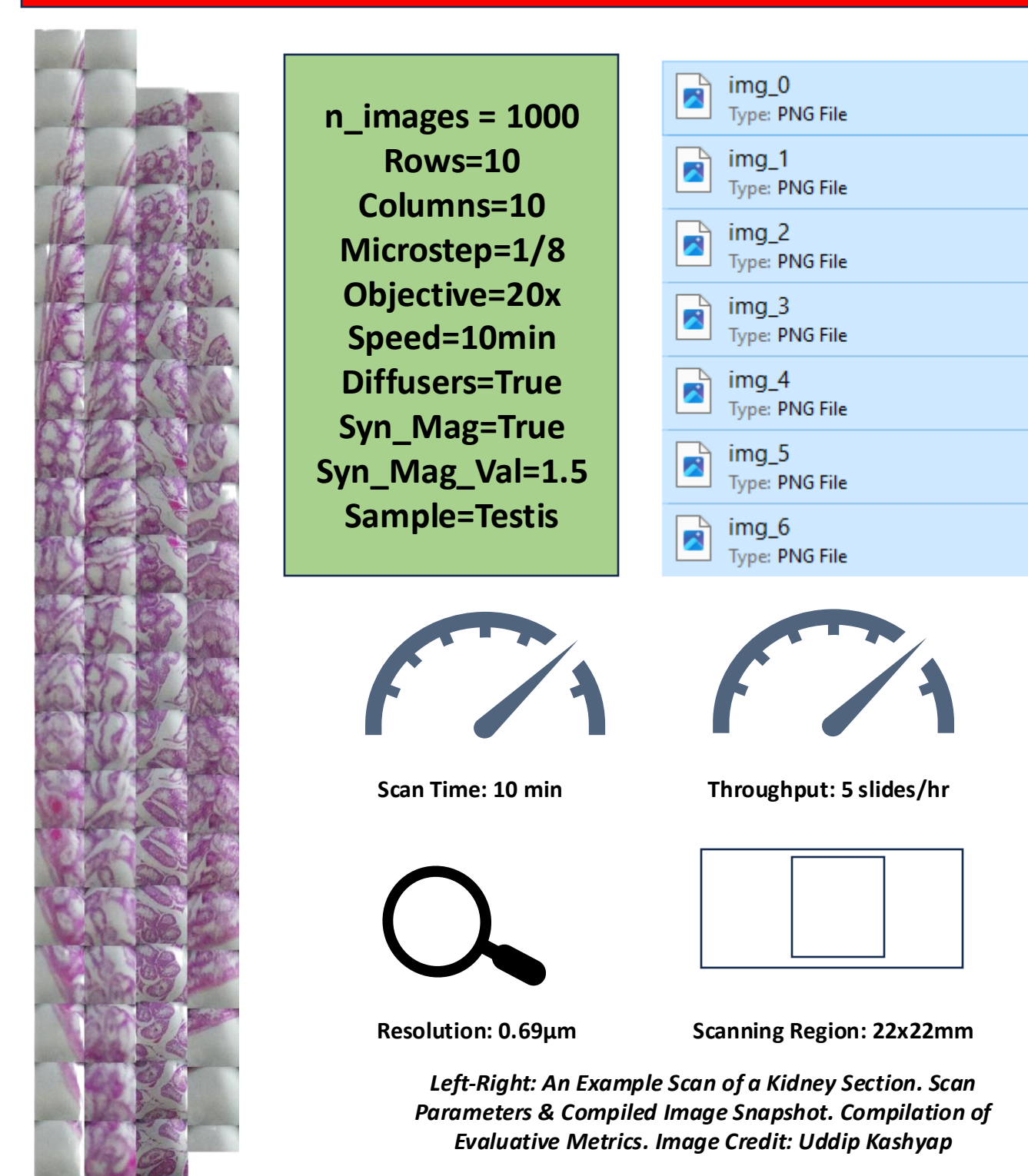
Front, Right Side, and Top views of the initial prototype. Image Credit: Uddip Kashyap

### SECONDARY PROTOTYPE



Front, Right Side, and Top views of the reproducible prototype. Image Credit: Uddip Kashyap

### SCAN RESULTS



Left-Right: An Example Scan of a Kidney Section. Scan Parameters & Compiled Image Snapshot. Compilation of Evaluative Metrics. Image Credit: Uddip Kashyap

## METHODOLOGY: AI

### Generator

