

Image Enhancement with Super-Resolution and Controlled Noise

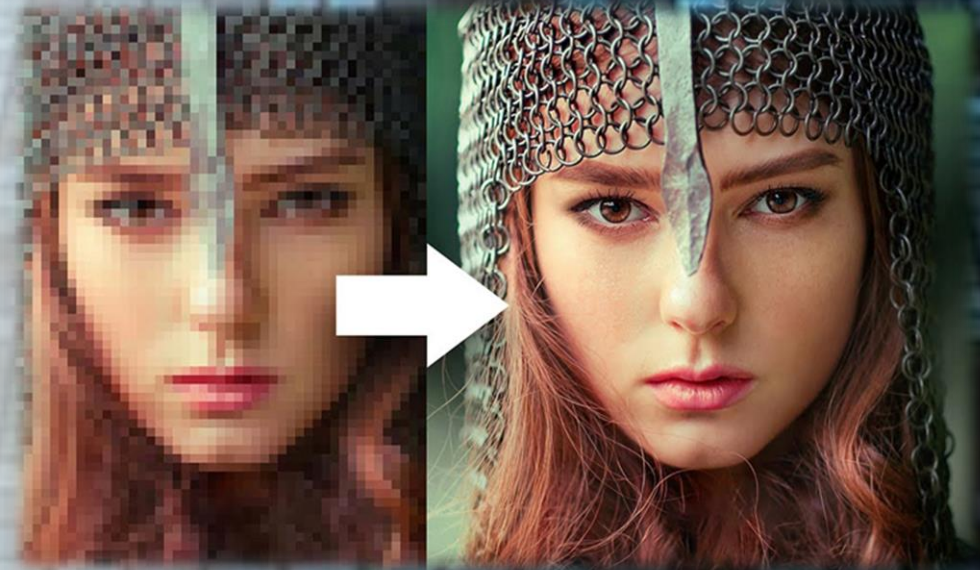
24-1-R-12

Students: Liza Shvachka

Avishai Hershkovitz

Supervisor: Dr. Renata Avros

Advisor: Prof. Zeev Volkovich



Introduction the problem

- Existing models rely on clean images for restoration, which are sometimes unavailable.
- Traditional models require deep understanding.
- Dealing with the real world where images are varied, including different exposure levels.

Importance of Image Restoration

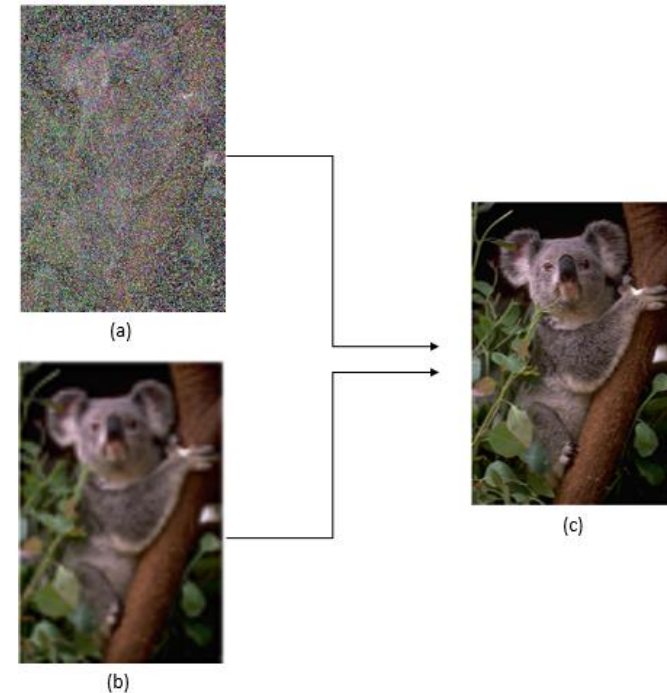
Image is a visual representation capturing scenes, and objects, essential for conveying ideas and experiences.

- Enhance clarity, detail, and overall quality.
- Improve visual understanding and interpretation.
- Correct imperfections for better usability.

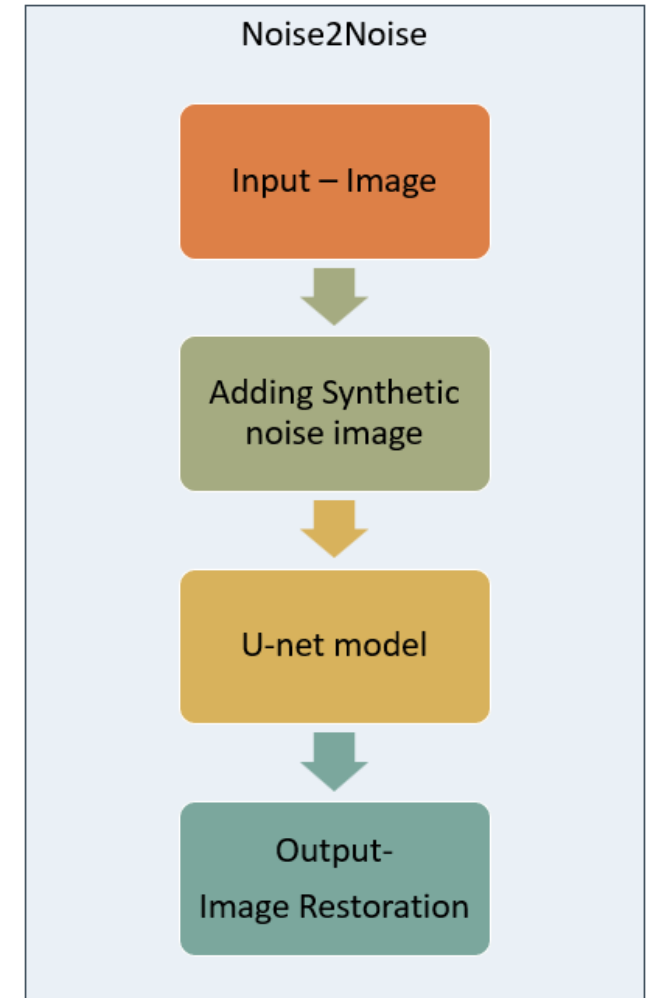
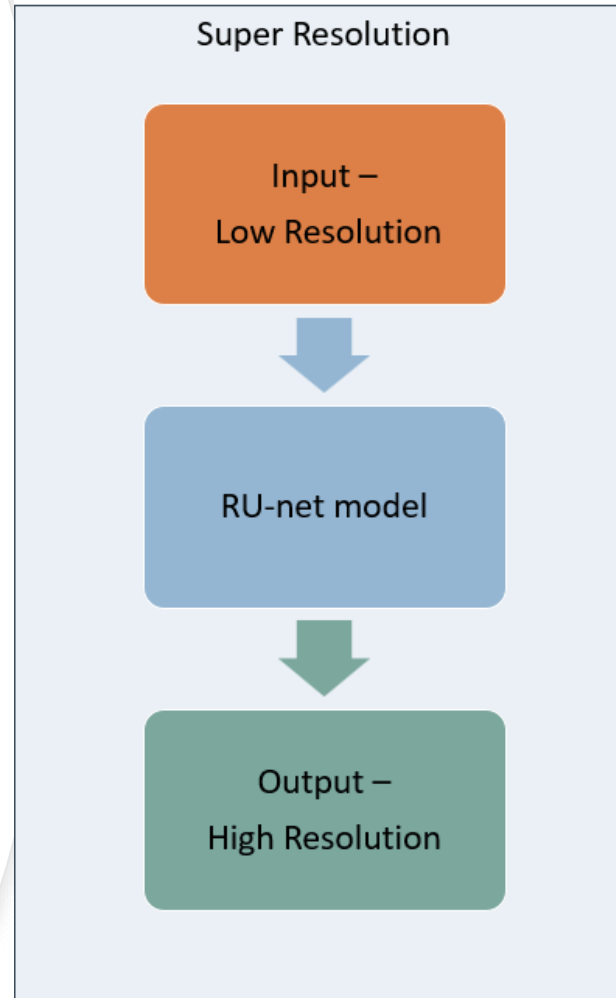


Suggested Solution

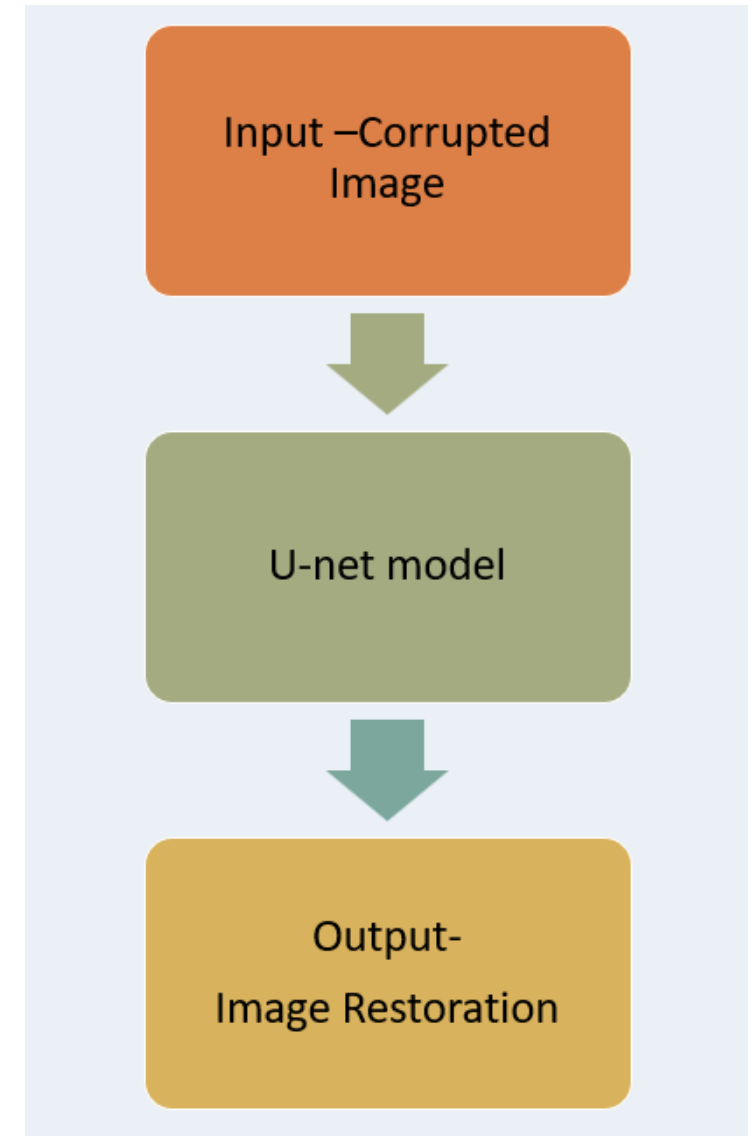
- Utilize Deep learning model Noise2Noise employing U-net.
- Restoring images based only on corrupted images.
- Implement Super Resolution Deep learning model using RU-net.



Flow Chart

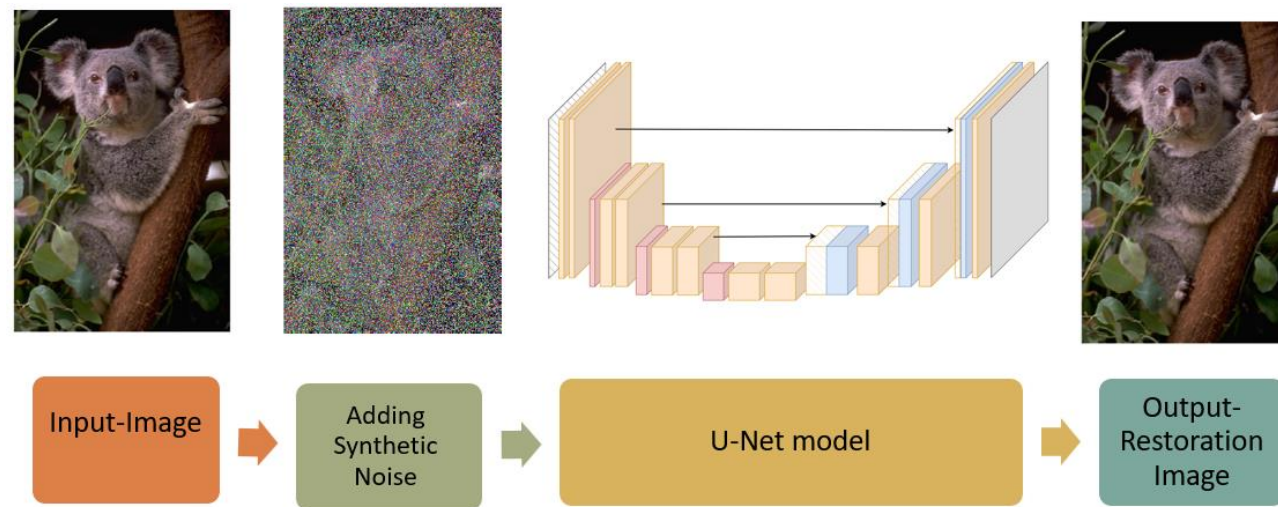


Noise2Noise

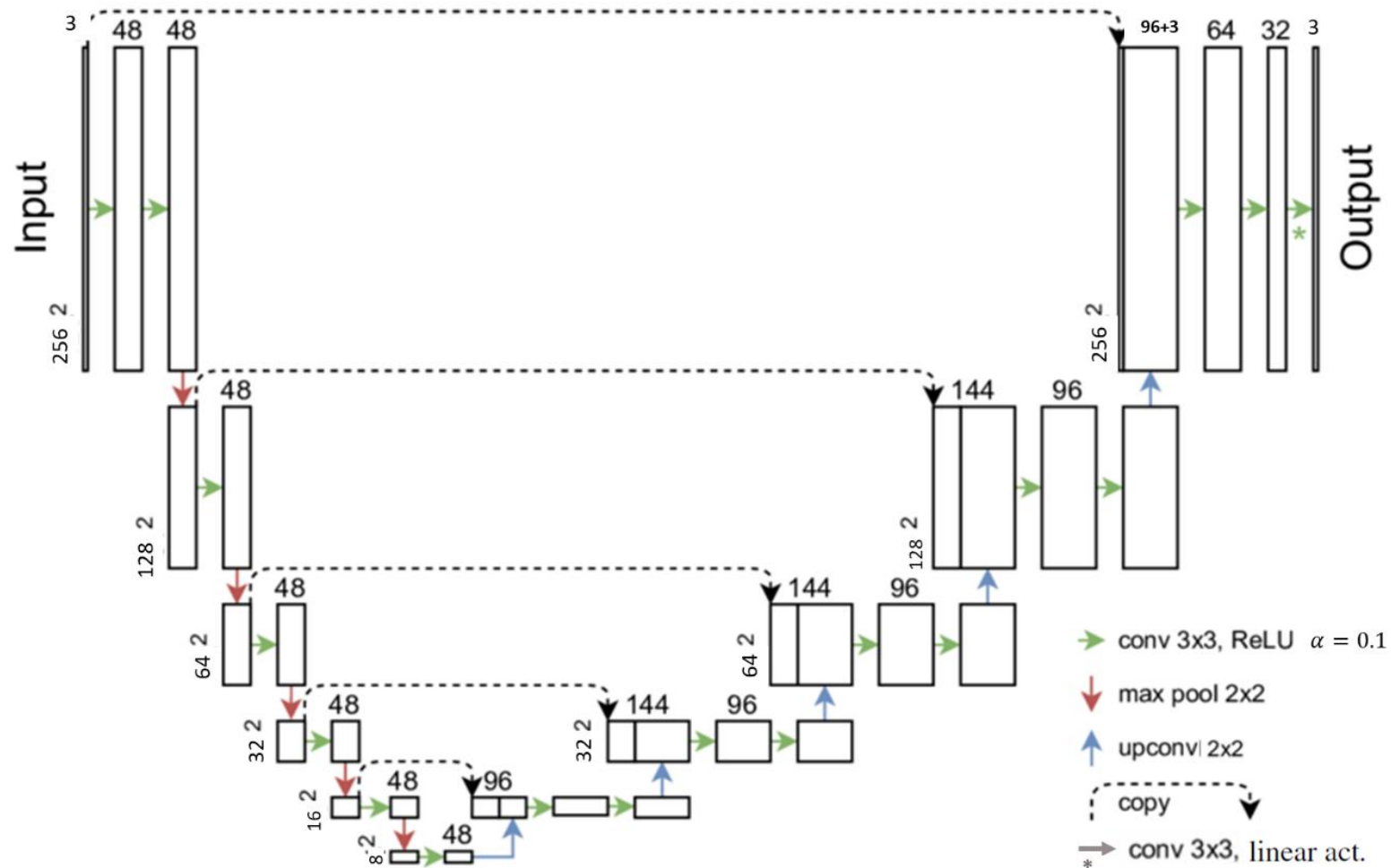


Noise2Noise Train Process

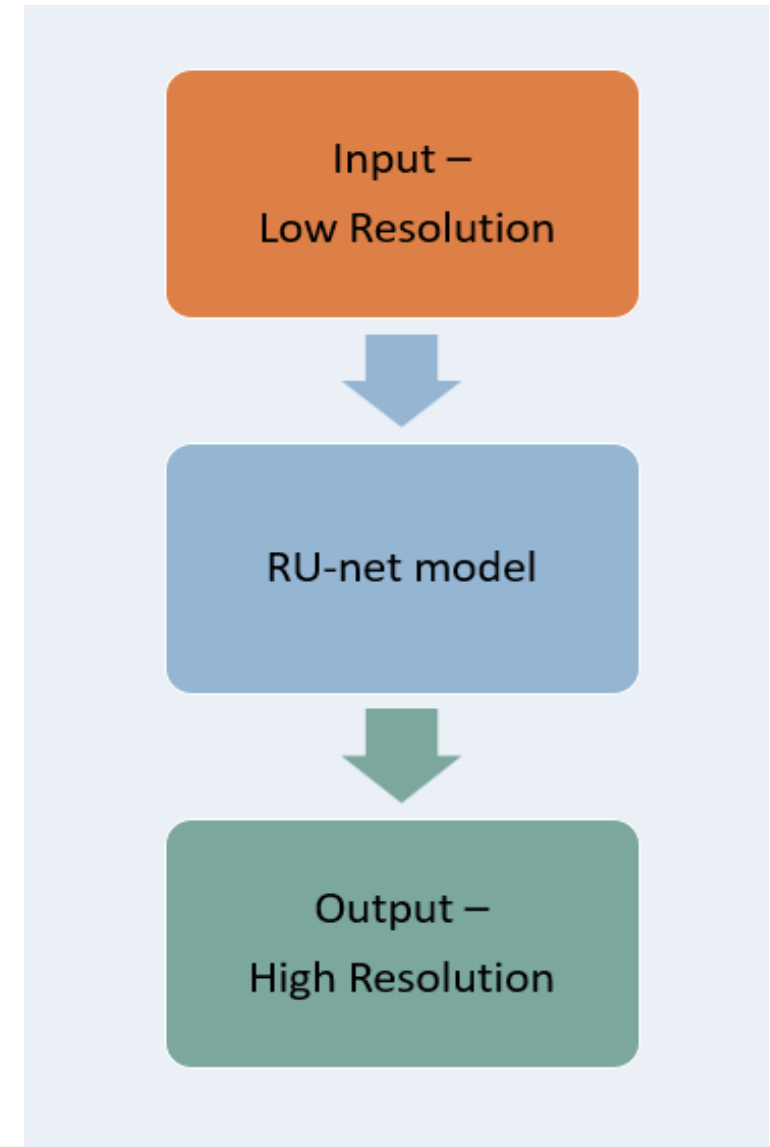
- Input: Image from various sources.
- Adding **Synthetic Noise**: Simulating real-world conditions.
- **U-net** Model Training: Iteratively mapping noisy to clean images.
- Output: Image Restoration with reduced noise.



Noise2Noise - *U-net* Phase

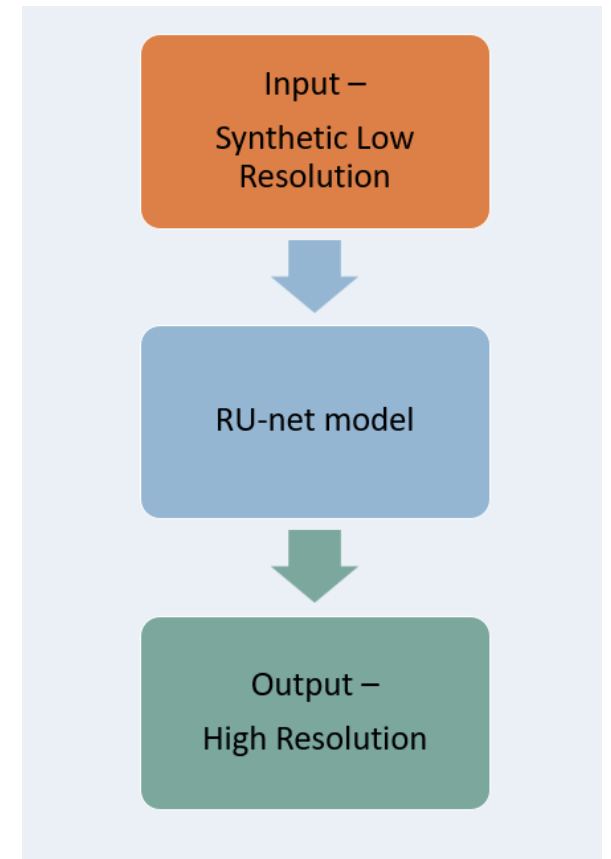


Super Resolution

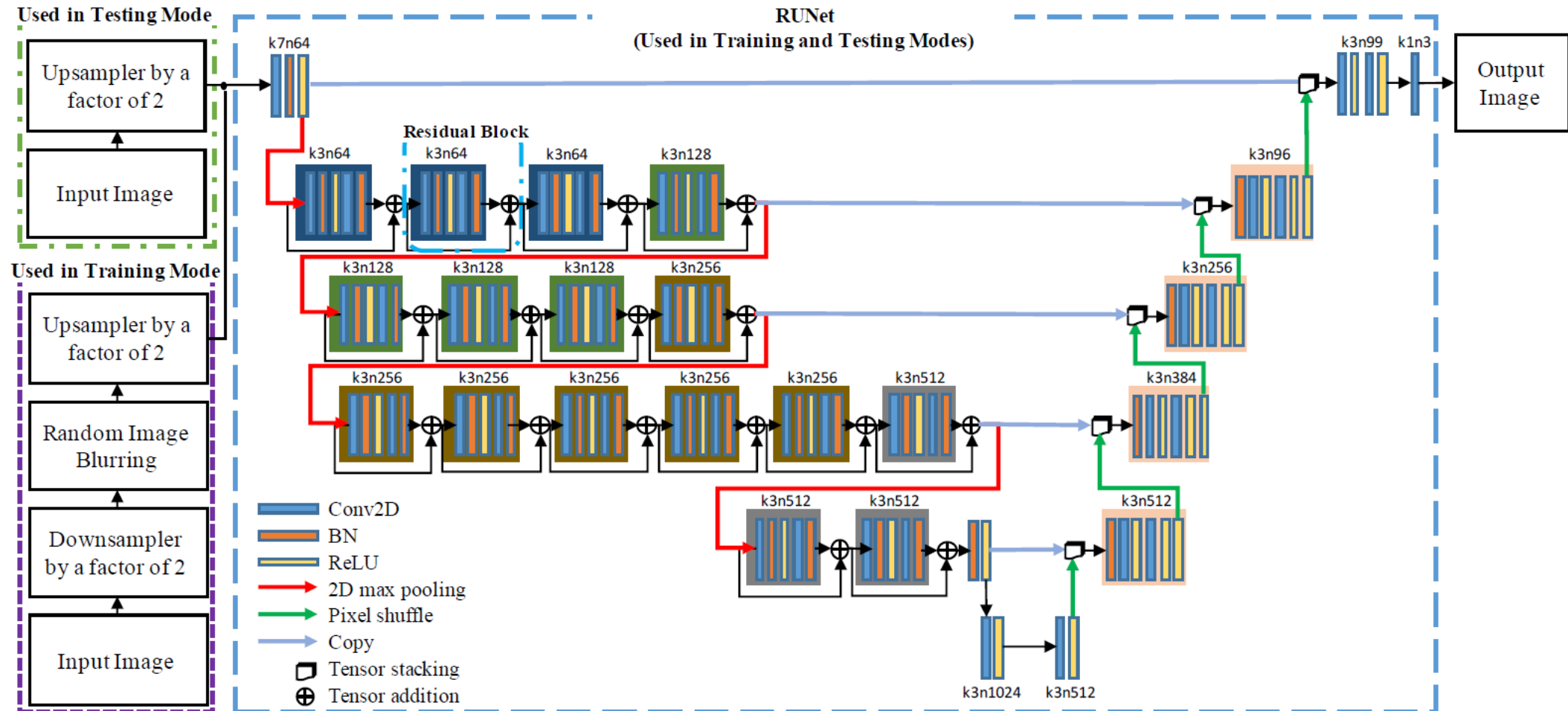


Super Resolution Train process

- Synthetic **Low-Resolution** input: High-resolution images, down-sampled, blurred, up-sampled.
- **RU-net** Model: A robust U-net architecture.
- Output: High-resolution images and visual enhancement.



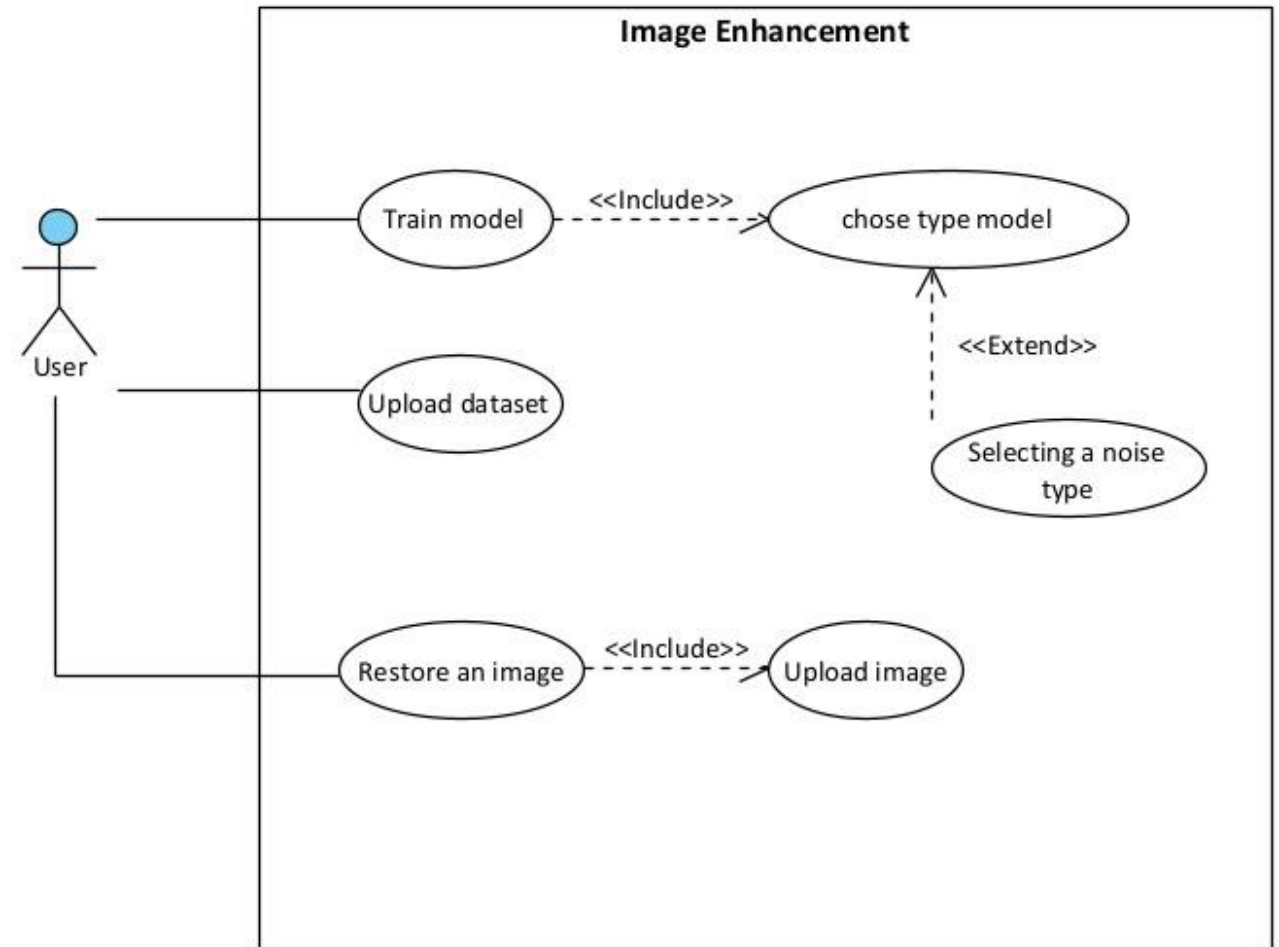
Super-Resolution - RU-net Phase



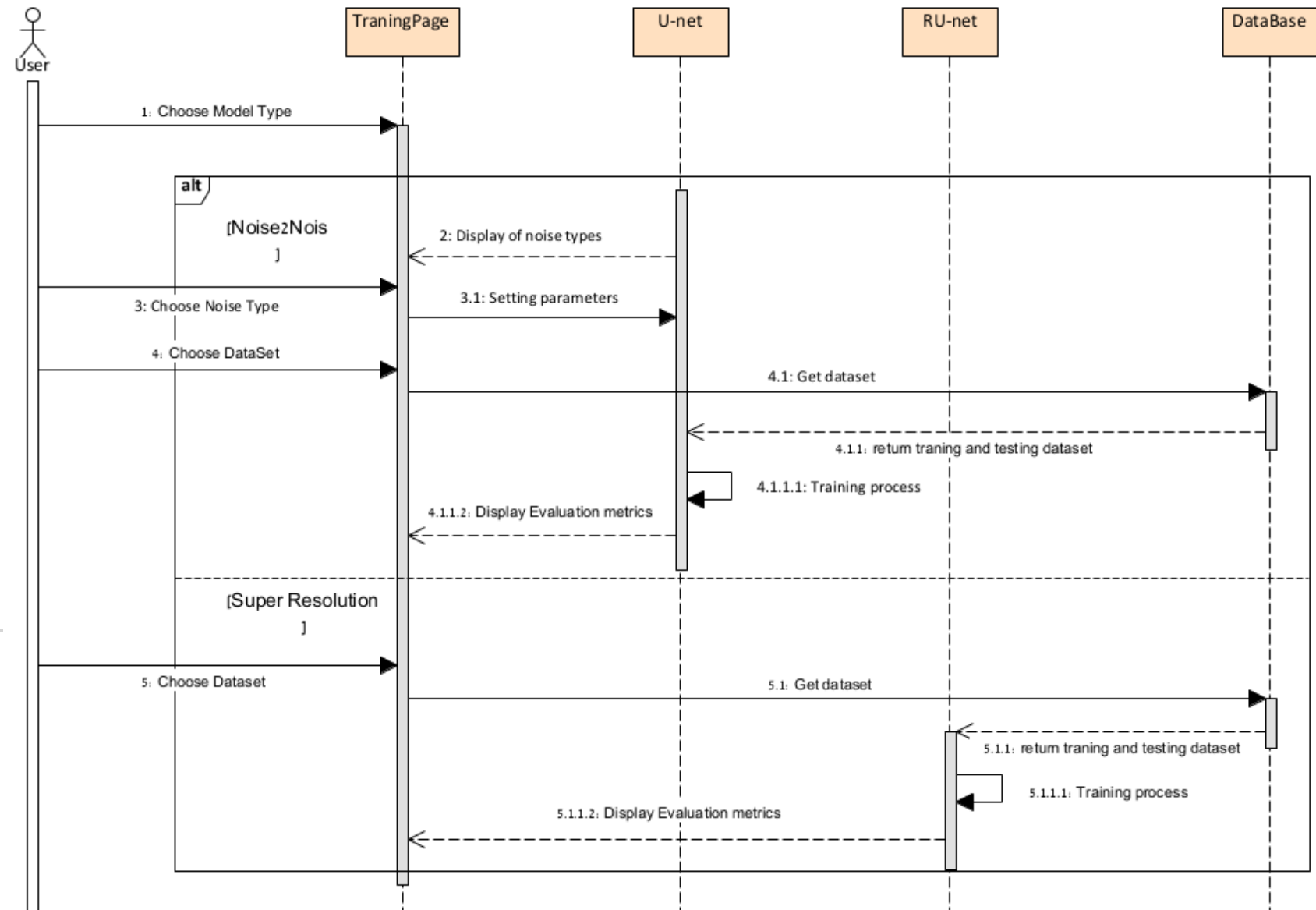
Expected Achievements

- Improve super-resolution and clean corrupted images techniques without explicit data.
- Compare the effectiveness of our models of reconstruction.
- Integrating both models into a unified model is planned.

Use Case




Sequence



GUI- Home page




GUI- Train Models page



Train models

Restore an image

Upload dataset



Train Models

Type of model:

Noise2Noise ▼

Super resolution

Noise2Noise

Dataset (train + test) :

dataset/data_set1 ▼

Type of noise (only for Noise2Noise):

Poisson ▼

Gaussian

Poisson


Multiplicative Bernoulli

Monte Carlo

Random text

Random-Valued

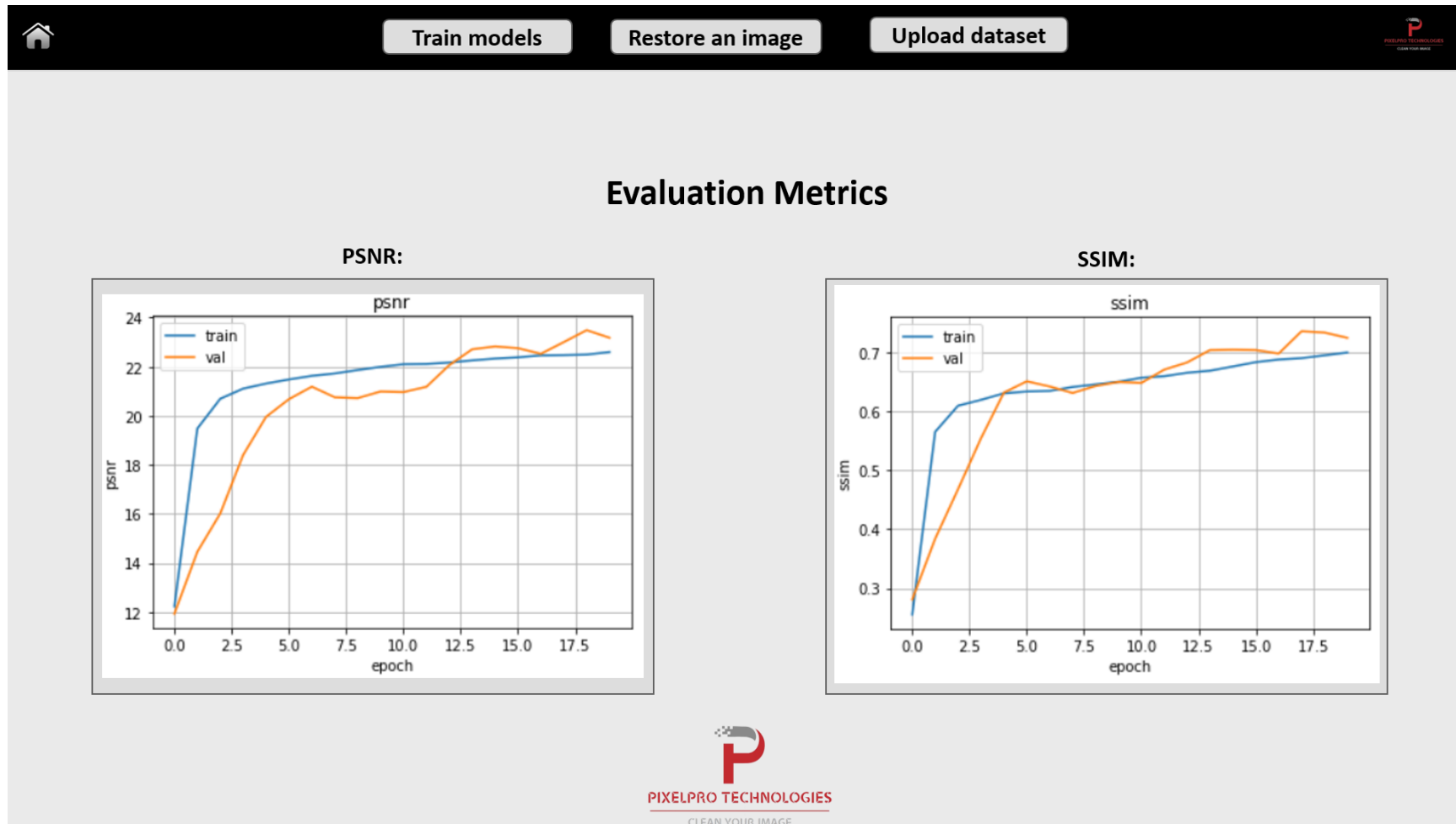
Train



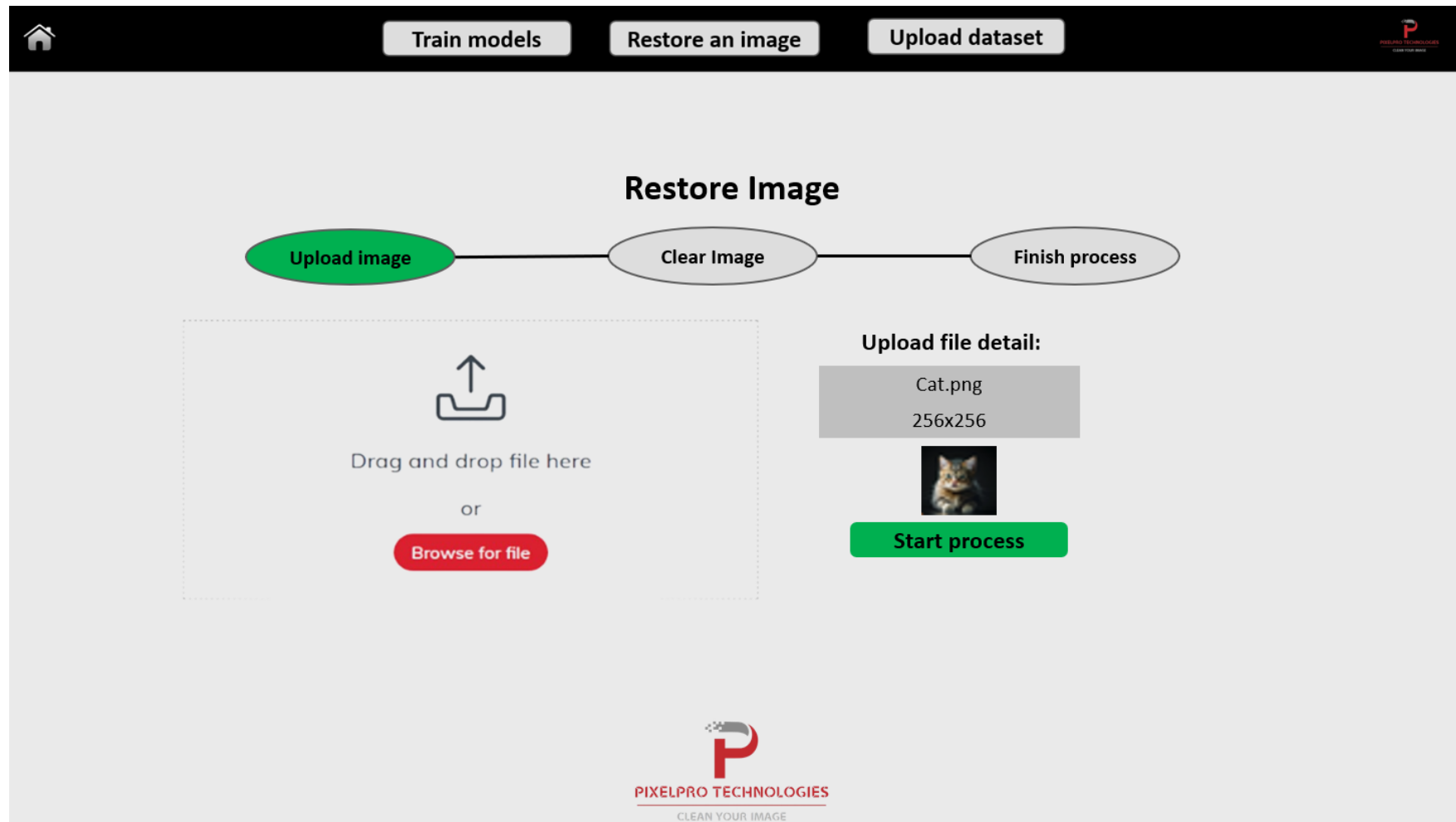
PIXELPRO TECHNOLOGIES

CLEAN YOUR IMAGE

GUI- Evolution metrics of train page



GUI- Restore Image page



Main Test process

Module	Test description	Excepted result
Home page	Click 'Home' button in navbar menu	Open 'Home' page
Home page	Click 'Train Models' in navbar menu	Open 'Train Models' page
Home page	Click 'Restore an image' button in navbar menu	Open 'Restore an image' page
Home page	Click 'Upload dataset' button in navbar menu	Open 'Upload dataset' page
Train Models window>Type of Model	Click on combo box 'Type of model' and choose 'Super resolution'	Option 'Super resolution' should be chosen.
Train Models window ->Type of model	Click on combo box 'Type of model' and choose 'Noise2Noise'	Option 'Noise2Noise' should be chosen
Train Models window ->Type of model (Noise2Noise)->Type of noise	Click on combo box 'Type of noise' and choose 'Gaussian'	Option 'Gaussian' should be chosen
Train Models window ->Train Process window	Click on 'Train' button	Open 'Train process' window with active training timer and current PSNR present window.
Restore an image window->Clear image	Click on button 'Start process'	Advance to step of Clear image processing with active precent of the clear
Restore an image window->Finish process	Click on button 'download'	Download the clean image to your computer
Upload dataset window	Click on 'Import dataset' button	An alert in green will pop up with message 'Dataset uploaded'



*Thanks for your
attention!*

Liza Shvachka
Avishai HersHKovitz

