

# Research Presentation

## Median Filtered Image Quality Enhancement and Anti-Forensics via Variational Deconvolution **and** **U-Net Architecture**



Presented By:

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B.Tech final yr.

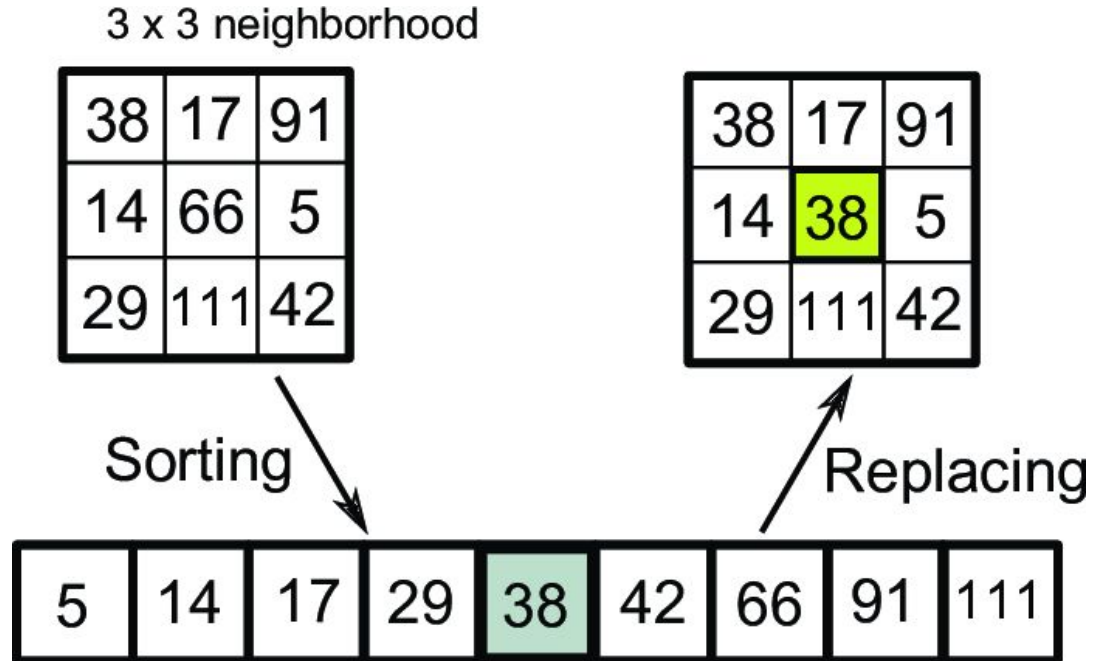
Supervised By:

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# Introduction

## Median Filtering

- Low Pass Filtering
- Non - Linear
- Smoothing Tool
- De-Noising Tool
- Salt & Pepper Noise



# Why Author targeted the Median Filtering

- Convicts uses Median Filtering to Hide Traces after tampering image

Original image

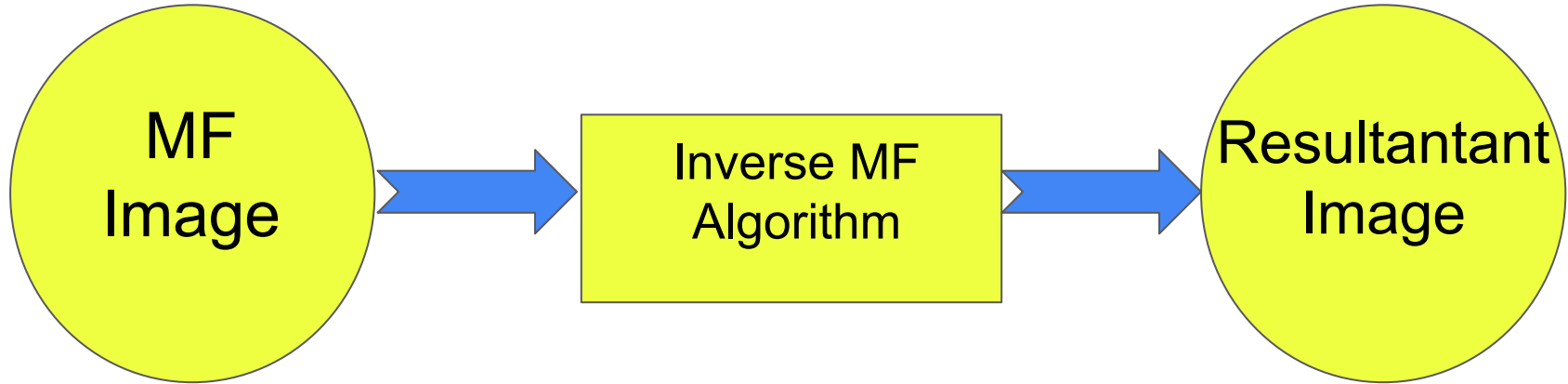


Filtered image



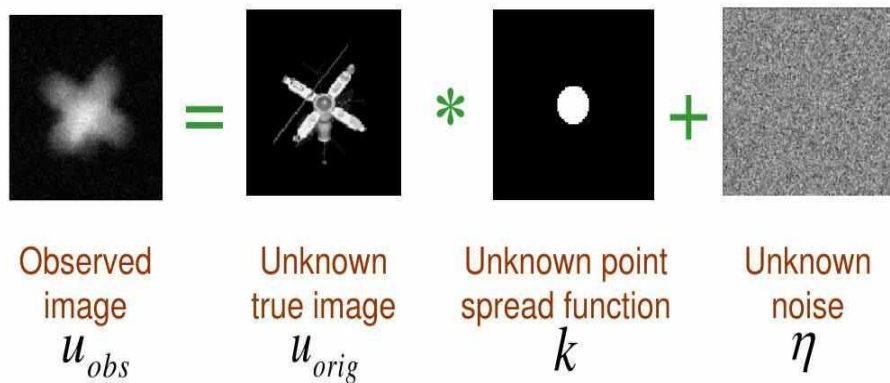
# Goal

- To deduce an Anti-Median Filtering Technique



# Proposed Solutions:

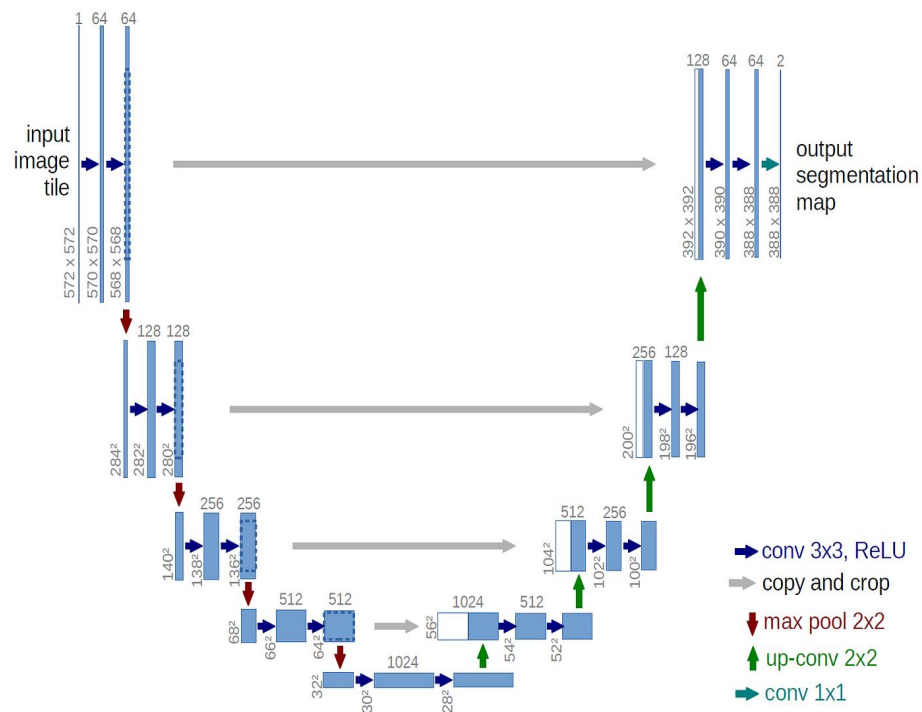
## Blind Deconvolution



Goal: Given  $u_{obs}$ , recover both  $u_{orig}$  and  $k$

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## U-Net Architecture



# Blind Deconvolution

“blind” = incomplete knowledge of kernel.

Proposed method for deconvolution:

$$\kappa = \frac{\Gamma(5/\beta)\Gamma(1/\beta)}{\Gamma(3/\beta)^2}$$

$$\tilde{\mathbf{x}} = \arg \min_{\mathbf{u}} \left( \frac{\lambda}{2} \left( \|\mathbf{K}\mathbf{u} - \mathbf{y}\|_2^2 + \omega \|\mathbf{u} - \mathbf{y}\|_2^2 \right) + \sum_{j=1}^J \left\| \frac{\mathbf{F}^j \mathbf{u}}{\alpha_j} \right\|_{\beta_j}^{\beta_j} \right).$$

$\mathbf{u}$  = MF img. (Matrix)

$\mathbf{K}$  = filter (approximated)

$\lambda, \beta, \alpha, \omega$  = approximation parameters obtained from other statistical equations.

# Blind Deconvolution

**Blur Model:** The observed (blurred) image  $g$  is typically modeled as:

$$g = h * f + n$$

- $h$ : Unknown blur kernel (PSF)
- $f$ : Original sharp image
- $n$ : Noise
- $*$ : Convolution operation

**Objective:** Estimate both  $f$  and  $h$  from the observed image  $g$ .

## Steps in Blind Deconvolution:

1. **Initial Estimation:** Start with an initial guess for the sharp image and blur kernel.
2. **Iterative Refinement:** Use iterative algorithms to refine the estimates of  $f$  and  $h$ .
3. **Regularization:** Apply constraints or regularization terms to ensure stable and realistic solutions, reducing the impact of noise.

# U-Net Architecture

## Brief Description:

- U-Net is a **Convolutional Neural Network (CNN)** designed for **image segmentation** and **restoration**.
- Features a **U-shaped structure** with an **encoder-decoder** design and **skip connections**.

## Key Benefits:

- **Retains Fine Details:** Skip connections prevent information loss.
- **Handles Complex Degradations:** Learns directly from data without estimating blur kernels.
- **Versatile:** Effective for tasks like **denoising**, **deblurring**, and **super-resolution**.



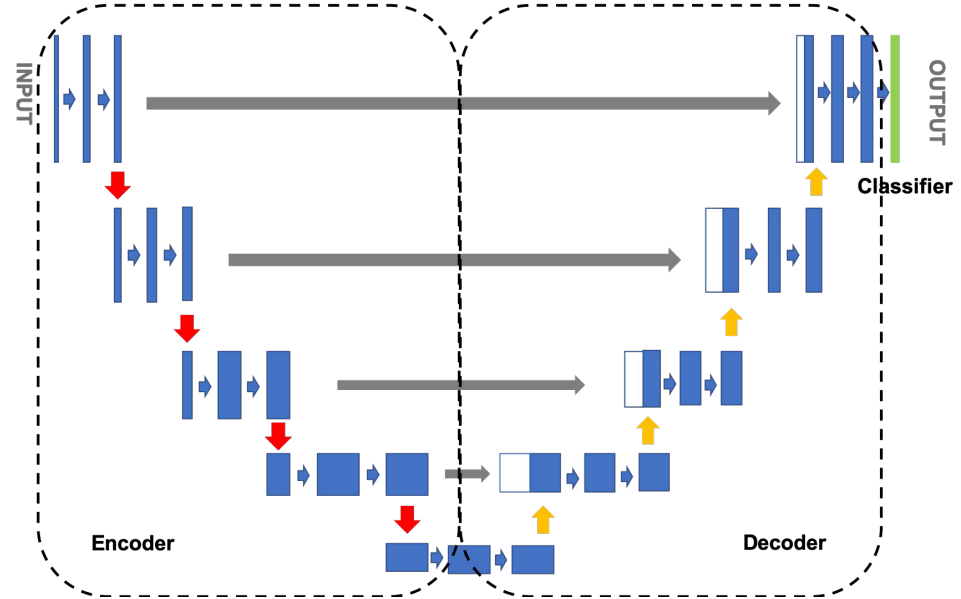
# U-Net Workflow

## Step-by-Step Workflow:

- **Input:** Blurred image.
- **Encoder:** Convolution + ReLU → Max Pooling (downsampling).
- **Bottleneck:** Convolutional layer (deep features).
- **Decoder:** Upsampling → Convolution → Skip Connection.
- **Output:** Restored (deblurred) image.

## Key Components:

- **Encoder:** Captures high-level features using convolution and downsampling (max pooling).
- **Bottleneck:** Connects the encoder and decoder, capturing deeper image representations.
- **Decoder:** Reconstructs the image using upsampling and feature concatenation from skip connections.
- **Skip Connections:** Bridge corresponding encoder-decoder layers to retain spatial details.



# Dataset & Processing Pipeline

Original dataset: **Pokemon-ds** [link](#)

# Images: 1538 (8-bit 512x512)

RGBA- ds



Grayscale-ds



Median Filtered ds



# Deconvolution Results (proposed by Wei Fan et. al.)

Median-Filtered Image



Deconvoluted Image



# Results Comparison

Variational Deconvolution

**Mse\_DC: 29.94275665283203**

Min\_Mse\_DC: 1.389920711517334

Max\_Mse\_DC: 633.2599487304688

U-Net with BCE Loss

Train Loss: 0.1541

**Validation Loss: 0.1606**

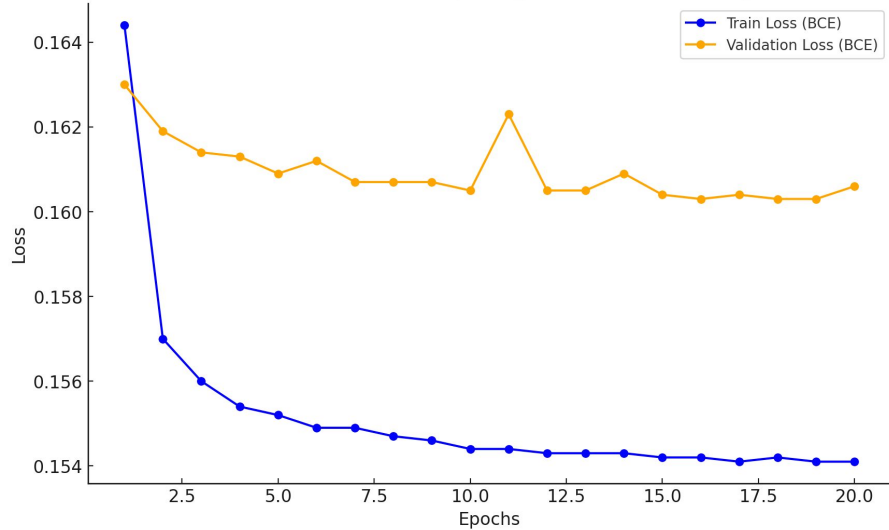
U-Net with MSE Loss

Train Loss: 0.0011

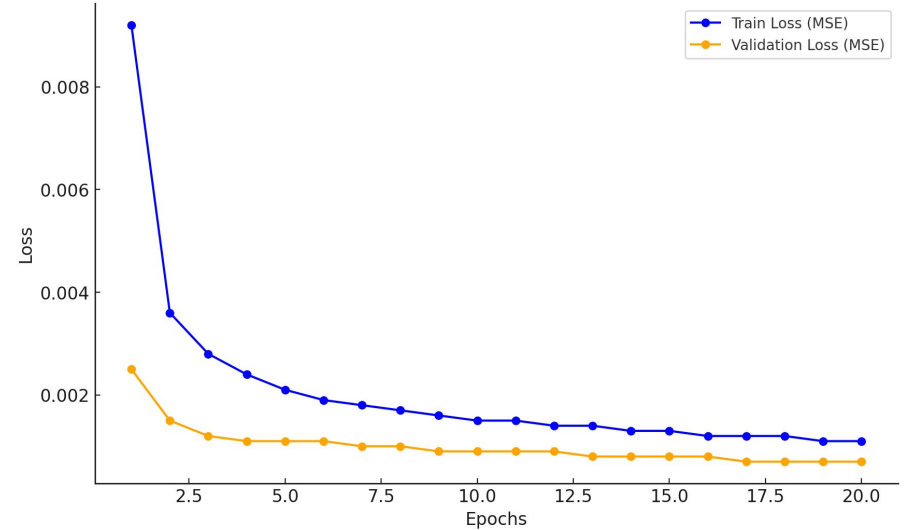
**Validation Loss: 0.0007**

# U-Net Loss Function Comparison

Model Loss Metrics (Using BCE Loss Function)



Model Loss Metrics (Using MSE Loss Function)



**Results:** U-Net performs better with MSE loss Function

# U-Net Results (Self - Proposed)

Median-Filtered Image



U-Net Generated Image



## References:

- <https://ieeexplore.ieee.org/document/7027174> (original Paper Link)
- [https://github.com/Akshat6133/De-Median\\_Filtering\\_using\\_DeConvolution\\_and\\_U-Net](https://github.com/Akshat6133/De-Median_Filtering_using_DeConvolution_and_U-Net)
- [https://lindevs.com/uploads/posts/content/2021/06/image\\_filtering\\_using\\_median\\_filter\\_and\\_opencv.png?v=1680423474](https://lindevs.com/uploads/posts/content/2021/06/image_filtering_using_median_filter_and_opencv.png?v=1680423474)
- [https://miro.medium.com/max/1400/1\\*lvXoKMH0PJMKpKK7keZMEA.png](https://miro.medium.com/max/1400/1*lvXoKMH0PJMKpKK7keZMEA.png)

Let's Move to Coding(600 lines) part

Thank you