What does Mean Filter do to Pixel Values?

Step 1 — What happens in smoothing?

- Smoothing = Reducing sharp changes (reducing edges/noise).
- It works by averaging nearby pixels.
- So if one pixel is **very bright** (**high value**) and its neighbors are **darker** (**low values**), the mean filter will **reduce that pixel's value** to bring it **closer to the average**.
- Similarly, if one pixel is **very dark** (**low value**) and its neighbors are brighter, the filter will **increase that pixel's value**.

☐ Mean Filter = Balance between High and Low Pixels

Pixel Type

Effect after Mean Filter

Very Bright Pixel (near dark ones) Brightness **decreases**Very Dark Pixel (near bright ones) Brightness **increases**Flat/Uniform Area

Very little change (values already close)

☐ What does this mean for Edges?

- Edges are where **pixel values change sharply** (dark-to-bright or bright-to-dark).
- Mean filter **blurs** these edges by reducing the difference between edge pixels and their neighbors.
- This is why smoothing with a mean filter can **reduce edge sharpness** (edges become softer).

Filter

In **digital image processing**, a **filter** is used to modify or enhance an image by processing pixel values based on a specific rule or pattern.

Example:

A **blur filter** smooths an image by averaging the pixel values with its neighboring pixels.

Original:

```
[10, 20, 30]
[40, 50, 60]
[70, 80, 90]
```

Applying a 3x3 blur filter:

• New center value = $(10 + 20 + 30 + 40 + 50 + 60 + 70 + 80 + 90) \div 9 = 50$

Result:

```
[10, 20, 30]
[40, 50, 60]
[70, 80, 90]
```

General Classification of Filters:

1. Linear Filter

A **linear filter** processes the image by applying a linear mathematical operation (like addition or multiplication) to the neighboring pixel values.

- The output is a weighted sum of the input pixel values.
- Examples: Mean filter, Gaussian filter, Laplacian filter

2. Nonlinear Filter

A **nonlinear filter** processes the image using a nonlinear operation (like finding the maximum, minimum, or median) on neighboring pixel values.

• Examples: Median filter, Bilateral filter

In the **spatial domain** \rightarrow Convolution with a kernel (e.g., mean, Gaussian)

In the **frequency domain** → Multiplying the Fourier transform of the image with a filter (e.g., low-pass, high-pass)

Example:

- Gaussian Filter (spatial): Blurs the image by averaging neighboring pixels.
- Low-pass Filter (frequency): Removes high-frequency noise in the Fourier domain.

1. Blurring (Smoothing)

- Goal: Reduce noise and smooth the image.
- Method:
 - Spatial domain → Averaging or Gaussian filter (reduces pixel value variation).
 - \circ **Frequency domain** \to Low-pass filter (removes high-frequency components).

Effect: Loss of detail, reduced sharpness.

2. Sharpening

- Goal: Enhance edges and fine details.
- Method:
 - o **Spatial domain** → Laplacian or Sobel filter (increases edge contrast).
 - o **Frequency domain** → High-pass filter (retains high-frequency components).

Effect: Increased detail, better edge visibility.

Purpose of Sharpening Spatial Filters:

- Opposite of smoothing filters (which blur the image).
- Used to remove blurring and improve the visibility of edges and textures.
- Based on **first-order** and **second-order derivatives** (which measure how fast the pixel values change).

1. First-Order Derivative (Gradient-Based)

Measures the **rate of change** in pixel intensity — detects the presence of an edge.

Formula:

f'=f(x+1)-f(x)

where:

- f(x) = pixel value at position x
- f(x+1) = pixel value at next position

Properties:

| ☐ Zero in flat areas (no change in pixel intensity) | |
|---|--|
| \square Non-zero at the start of an edge (step change). | |
| ☐ Non-zero along a slope (gradual change). | |

First-Order Derivative in 2D (Edge Detection):

Common filters based on first-order derivatives:

- Sobel filter
- Prewitt filter
- Roberts filter

These filters calculate gradients in **both x and y directions** to detect edges.

2. Second-Order Derivative (Laplacian-Based)

Measures the **rate of change of the gradient** — highlights the exact location of edges.

Formula:

$$f''=f(x+1)+f(x-1)-2f(x)$$

Properties:

 \square Zero in flat areas (constant pixel intensity).

☐ Non-zero at the start **and end** of an edge (step change).

☐ Zero along gradual changes (ramps).

Second-Order Derivative in 2D (Edge Detection):

Common filters based on second-order derivatives:

- Laplacian filter
- LoG (Laplacian of Gaussian)

Summary:

- 1. First-Order Derivative \rightarrow Edge Detection
- First-order filters (like **Sobel** or **Prewitt**) detect where the edges are by finding areas with a **sharp change** in pixel values.
- Output: Highlights edges but may not be sharp enough.

Example Workflow:

- 1. Apply a **Sobel filter** \rightarrow Detect edges.
- 2. Apply a **Laplacian filter** \rightarrow Sharpen the edges.
- 3. Combine the results \rightarrow Enhanced, clear, and sharp edges.

2. Second-Order Derivative → Edge Sharpening

- Second-order filters (like **Laplacian**) enhance those detected edges by making them **sharper** and more defined.
- Output: Improves clarity and makes edges more prominent.

1. Flat Area (No Change in Brightness)

Example:

[10,10,10,10,10]

- All pixel values are the same \rightarrow No change
- No edge or slope → Nothing to detect

☐ First-order:

• $10-10=0 \rightarrow \text{No change} \rightarrow \text{Zero}$

☐ Second-order:

• $10+10-2(10)=20-20=0 \rightarrow \text{No change} \rightarrow \textbf{Zero}$

 \checkmark Both first and second order = **0** (because there's no change).

□ 2. Step (Sudden Change in Brightness)

Example:

[10,10,50,50,50]

- Jump from 10 to $50 \rightarrow \text{Sharp edge}$
- First-order will **detect** the edge

| • | Second-order will enhance the edge |
|--------|--|
| □ Firs | st-order: |
| • | 50 − 10 = $40 \rightarrow$ Detects edge \checkmark |
| | cond-order: |
| • | $50+10-2(10)=40$ → Enhances edge \checkmark |
| □ 3. | Ramp (Gradual Change in Brightness) |
| Examp | ple: |
| [10,20 | ,30,40,50] |
| • | Brightness increases gradually → Slope First-order detects the slope Second-order ignores it (because it's smooth) |
| □ Firs | st-order: |
| • | 20–10=10, 30–20=10, 40–30=10 → Constant slope \checkmark |
| | cond-order: |
| • | $30+10-2(20)=40-40=0 \rightarrow \text{No sudden change} \rightarrow \textbf{Zero}$ |
| Flat A | Area → No enhancement |
| • | No change → Both first and second derivatives are zero → No enhancement possible |

 $Ramp \rightarrow No enhancement$

• Gradual change \rightarrow First-order detects it, but second-order is zero \rightarrow No enhancement

Step → Enhancement happens!

Sudden change → First-order detects the edge → Second-order enhances it by increasing contrast → Enhancement happens

Example

First-Order Derivative → Edge Detection

• When you use a tool to **detect edges** in a photo (like "Find Edges" in Photoshop), it applies a first-order filter to identify the boundaries of objects.

Example: Highlighting the outline of a face or object.

Second-Order Derivative → Edge Enhancement

• When you use a tool to **sharpen** a photo, it applies a second-order filter to enhance the edges by increasing contrast.

Example: Making facial features sharper or improving text clarity in an image.