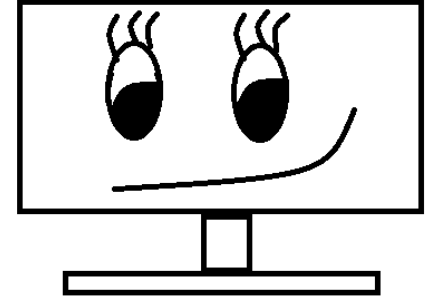


# Seneca



## **CVI620/ DPS920**

# **Introduction to Computer Vision**

## **Noise & Filtering**

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# Overview

- Noise
  - Gaussian
  - Impulsive (Salt & Pepper)
- Filtering
  - Linear Filtering
  - Nonlinear Filtering

# Noise [3]

- Noise: anything that degrades the ideal image
- Sources of noise:
  - The environment,
  - The imaging device,
  - Electrical interference,
  - The digitization process, and so on.
- Noise is additive and random:

$$\hat{I}(i, j) = I(i, j) + n(i, j)$$

# Gaussian Noise [3]

- A good approximation of real noise
- Modelled as a Gaussian (normal distribution with mean of 0)
- $n \sim N(\mu = 0, \sigma)$



# Impulsive noise- Salt and Pepper Noise

- Impulsive: noise peaks or spikes
- Salt & Pepper is a model often used for impulsive noise
- Random values of brightness (darker or lighter) in the image
- Let  $p$  be the probability of noise ( $0 \leq p \leq 1$ )
- And  $x$  and  $y$  two random numbers between 0 and 1

$$I_{noisy}(i, j) = \begin{cases} I_{min} + y(I_{max} - I_{min}) & \text{if } x < p \\ I_{in}(i, j) & \text{otherwise} \end{cases}$$

- In 8-bit images,  $I_{min}=0$  and  $I_{max}=255$



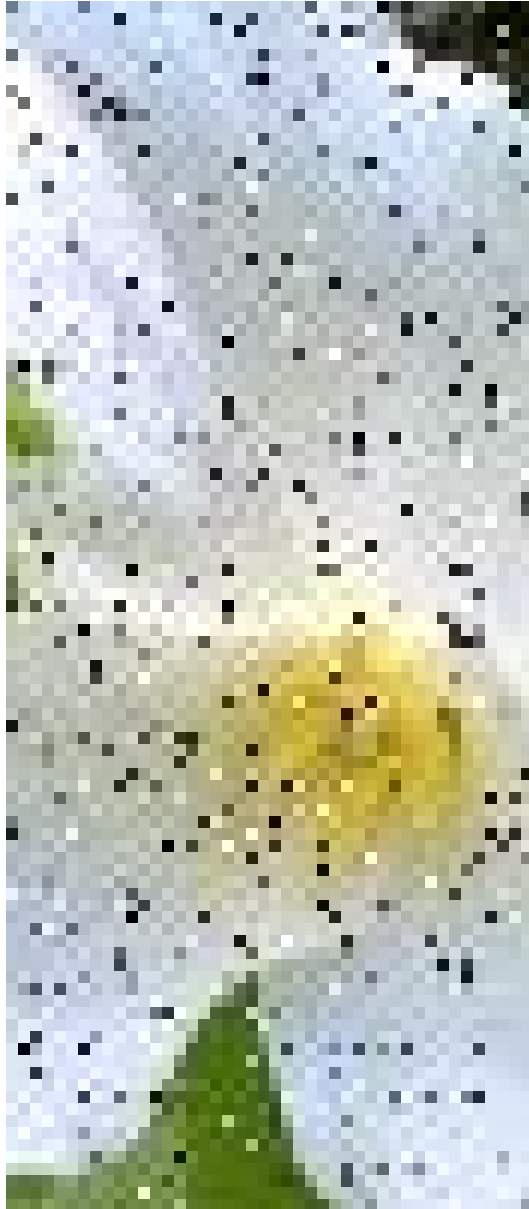
# Examples

$p = 0.1$



$p = 0.5$





# Correcting noise

- Observation: The image does not change sharply most of the time (low frequency), while noise is a sharp peak (high frequency)
- Therefore using the values of the neighbors, we can often lower the noise
- Take the average of the neighboring pixels
  - This is equivalent to low-pass filtering, which keeps the low frequency content and filters high frequency signals
- Disadvantage: This will reduce the sharpness of edges in the image

# Averaging

- The value at pixel (i, j) is calculated as the average of the pixels in its neighborhood
- Suitable for removing random noise, or smoothing

$j$  →

↓  $i$

62	79	23	119	120	105	4	0
10	10	9	62	12	78	34	0
10	58	197	46	46	0	0	48
176	135	5	188	191	68	0	49
2	1	1	29	26	37	0	77
0	89	144	147	187	102	62	208
255	252	0	166	123	62	0	31
166	63	127	17	1	0	99	30

$I_{in}$

				?			

$I_{out}$

5x5 neighborhood

$$\text{new value} = \frac{9 + 62 + \dots + 102 + 62}{25}$$



# Linear Filtering

# Linear filtering

- **Filtering:** an algorithm that starts with some image  $I_{\text{in}}(i, j)$  and computes a new image  $I_{\text{out}}(i, j)$  using a neighborhood operator
- **Kernel:** A template defining the neighborhood and the operator
- **Linear filter / linear kernel:** Values are calculated as a weighted sum of values in the neighborhood

$$I_{\text{out}}(i, j) = \sum_{x, y \in \text{Kernel}} k(x, y) \cdot I_{\text{in}}(i + x, j + y)$$

# Averaging- box kernel

- Averaging is equivalent to convolution with a box kernel

$i \downarrow$   $j \rightarrow$

62	79	23	119	120	105	4	0
10	10	9	62	12	78	34	0
10	58	197	46	46	0	0	48
176	135	5	188	191	68	0	49
2	1	1	29	26	37	0	77
0	89	144	147	187	102	62	208
255	252	0	166	123	62	0	31
166	63	127	17	1	0	99	30

$k = 1/25 \times$

1	1	1	1	1
1	1	1	1	1
1	1	<u>1</u>	1	1
1	1	1	1	1
1	1	1	1	1

5x5 (normalized) box kernel

- $I_{out} = I_{in} * k$

Convolution (\*) is a mathematical operation

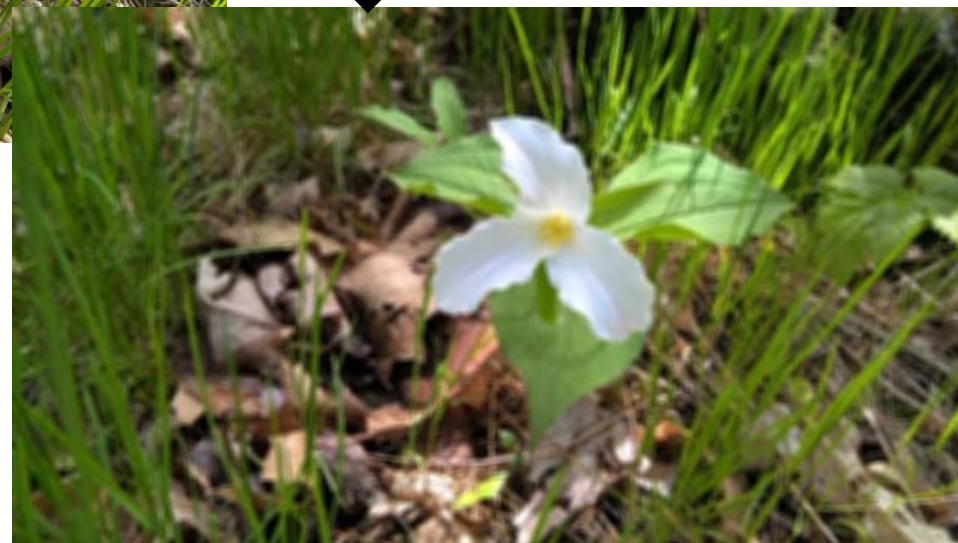
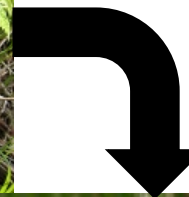
$k = 1/9 \times$

1	1	1
1	<u>1</u>	1
1	1	1

3x3 (normalized) box kernel



5x5 (normalized) box kernel



```
// Using this function  
blur(img, dst, cv::Size(5, 5));  
  
// Or use this function  
boxFilter(img, dst, CV_8U, cv::Size(5,5));  
  
// Or build a box kernel yourself and then filter  
Matx<float,5,5> myK= cv::Matx<float,5,5>::all(1 / 25.0f);  
filter2D(img, dst, CV_8U, myK);
```



# Examples

Salt & pepper noise with  $p = 0.1$



After 5x5 box filter





# Separable filtering

Some filters are separable into smaller filters. Applying smaller filters is faster (faster implementation).

For example:  
Convolving with

 $\frac{1}{25} \times$ 

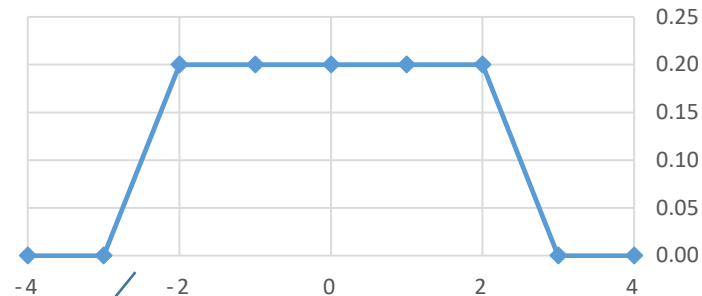
1	1	1	1	1
1	1	1	1	1
1	1	<u>1</u>	1	1
1	1	1	1	1
1	1	1	1	1

5x5 (normalized) box kernel

Is equivalent to  
convolving with

 $\frac{1}{5}$ 

1	1	<u>1</u>	1	1
---	---	----------	---	---



A low-pass filter!

And then  
convolving with

 $\frac{1}{5}$ 

1
1
<u>1</u>
1
1

# Gaussian Filter (smoothing)

- The Gaussian Filter (2-D bell curve) is separable
- It can be applied by first convolving with a 1D Gaussian Filter horizontally and then vertically

- ```
cv::Mat cv::getGaussianKernel(  
    int          ksize,          // Kernel size  
    double       sigma,          // Gaussian half-width  
    int          ktype = CV_32F  // Type for filter coefficients  
);
```

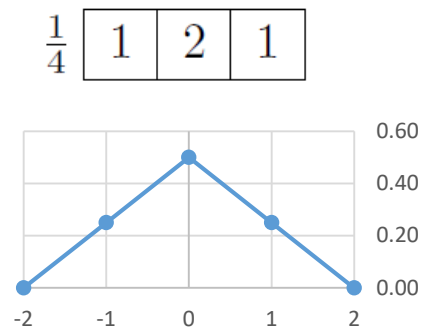
- The coefficients are computed from 
$$k_i = \alpha \cdot e^{-\frac{(i-(ksize-1)/2)^2}{(2\sigma)^2}}$$

- It can be applied using `sepfiler2D` (instead of `filter2D`)

# Bilinear kernel

- Also smoothing (removing noise)
- Equivalent to convolving with two separable 'tent' functions
- Example: 3x3 bilinear kernel:

1-D Tent kernel:



2-D Bilinear kernel:

$\frac{1}{16}$

|   |   |   |
|---|---|---|
| 1 | 2 | 1 |
| 2 | 4 | 2 |
| 1 | 2 | 1 |

# Examples

$p = 0.1$



After 3x3 bilinear filter



# Nonlinear Filtering



# Nonlinear filter

- The output pixel value is not a linear function of pixel values in the input
- Example: Median Filter
- The output value is the median of the pixels in the neighborhood

```
void cv::medianBlur(  
    cv::InputArray  src,           // Input image  
    cv::OutputArray dst,          // Result image  
    int             ksize         // Kernel size  
);
```

# Examples

$p = 0.1$



After 3x3 median filter



# References

- [1] Computer Vision: Algorithms and Applications, R. Szeliski  
(<http://szeliski.org/Book>)
- [2] Learning OpenCV 3, A. Kaehler & G. Bradski
  - Available online through Safari Books, Seneca libraries
  - [https://senecacollege-primo.hosted.exlibrisgroup.com/primo-explore/fulldisplay?docid=01SENC\\_ALMA5153244920003226&context=L&vid=01SENC&search\\_scope=default\\_scope&tab=default\\_tab&lang=en\\_US](https://senecacollege-primo.hosted.exlibrisgroup.com/primo-explore/fulldisplay?docid=01SENC_ALMA5153244920003226&context=L&vid=01SENC&search_scope=default_scope&tab=default_tab&lang=en_US)
- [3] Practical introduction to Computer Vision with OpenCV, Kenneth Dawson-Howe
  - Available through Seneca libraries
  - [https://senecacollege-primo.hosted.exlibrisgroup.com/primo-explore/fulldisplay?docid=01SENC\\_ALMA5142810950003226&context=L&vid=01SENC&search\\_scope=default\\_scope&tab=default\\_tab&lang=en\\_US](https://senecacollege-primo.hosted.exlibrisgroup.com/primo-explore/fulldisplay?docid=01SENC_ALMA5142810950003226&context=L&vid=01SENC&search_scope=default_scope&tab=default_tab&lang=en_US)