
Edge Detection via Objective functions

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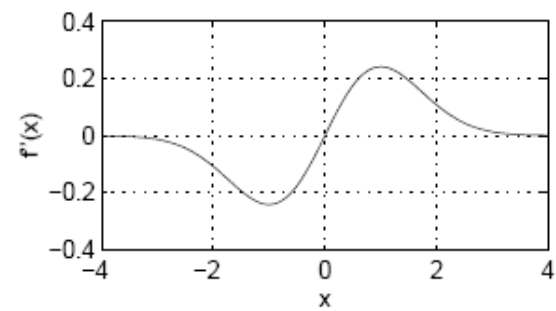
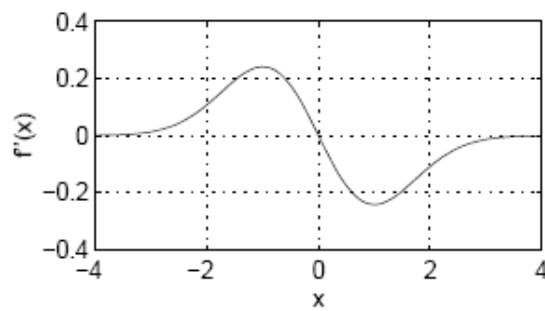
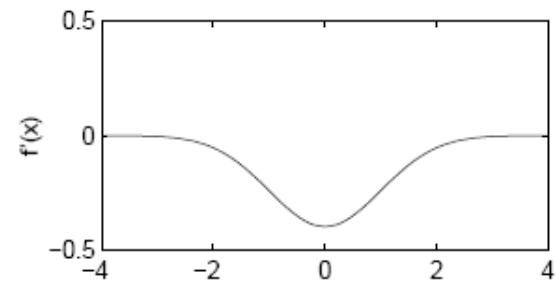
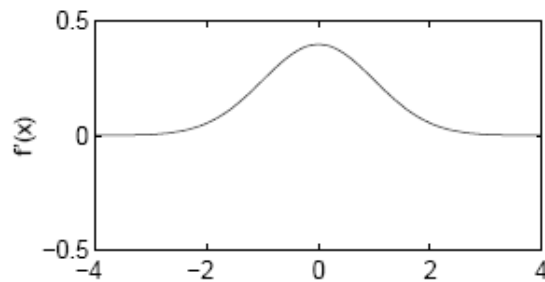
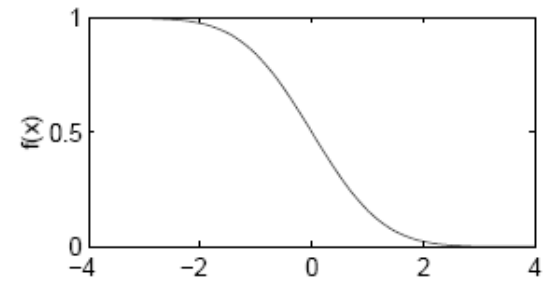
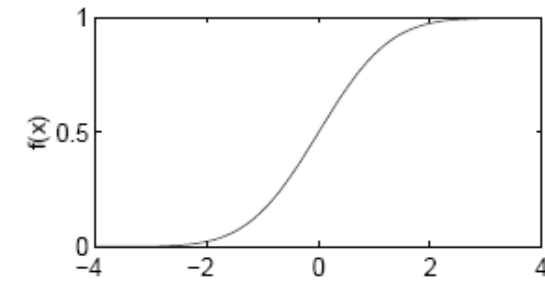
Edge Detection – a quick recap!

- Much of the “information” in an image is in the edges
- Edge detection is usually for the purpose of subsequent computer processing
- Examples : Image segmentation, object recognition, object detection
- General framework :

$$\underbrace{f(x, y)}_{\text{image}} \rightarrow \boxed{\text{edge detection system}} \rightarrow \underbrace{e(x, y)}_{\text{edge map}} = \begin{cases} 1, & (x, y) \in \text{edge} \\ 0, & \text{otherwise.} \end{cases}$$

$$\underbrace{f(x, y)}_{\text{image}} \rightarrow \boxed{\text{edge enhancer}} \rightarrow \boxed{\text{edge detector}} \rightarrow \boxed{\text{thinning}} \rightarrow \underbrace{e(x, y)}_{\text{edge map}}$$

Concept



- Approximate the derivative by finite differences

$$f'_a(x)|_{x=n\Delta_x} \approx \frac{f[n] - f[n-1]}{\Delta_x} \quad \text{left difference}$$
$$\frac{f[n+1] - f[n]}{\Delta_x} \quad \text{right difference}$$
$$\frac{f[n+1] - f[n-1]}{2\Delta_x} \quad \text{central difference}$$

A Classical method - Sobel

- Sobel's method :

Step 1: Utilizes masks S_x and S_y to obtain edge intensities E_x and E_y

Step 2: Thresholds the edge map to detect edge points.

Disadvantage : Edges detected are thicker than actual edges.

$$|\nabla f(x, y)| \approx \sqrt{(f_x[n, m])^2 + (f_y[n, m])^2}$$

$$f_x[n, m] \triangleq f[n, m] ** h_x[n, m] \text{ and } f_y[n, m] \triangleq f[n, m] ** h_y[n, m]$$

-1	0	1
-2	0	2
-1	0	1

S_x

1	2	1
0	0	0
-1	-2	-1

S_y

Canny

■ Canny's Method :

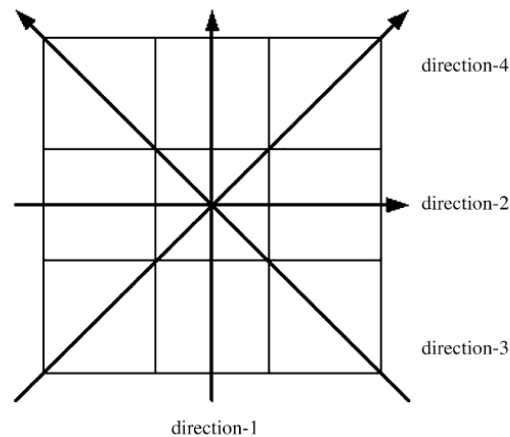
Step 1: Apply a Gaussian filter to the image

Step 2: Gradient of the smoothed image is computed in x and y directions

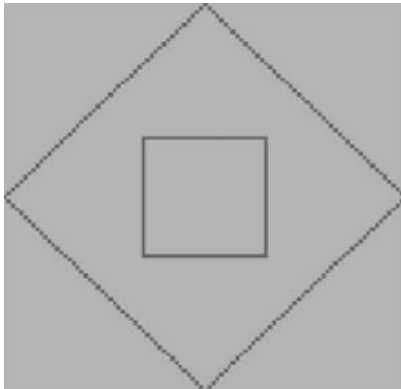
Step 3: Edge and Direction maps are generated

Step 4: Non Maxima Suppression (NMS) is performed

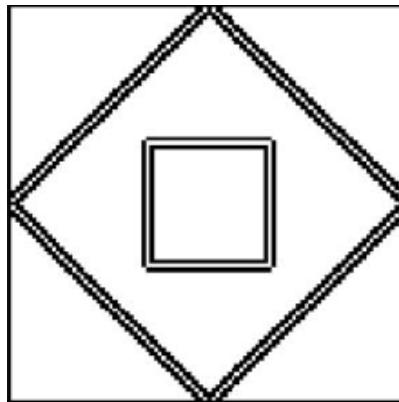
Advantage over Sobel : Thinner edges



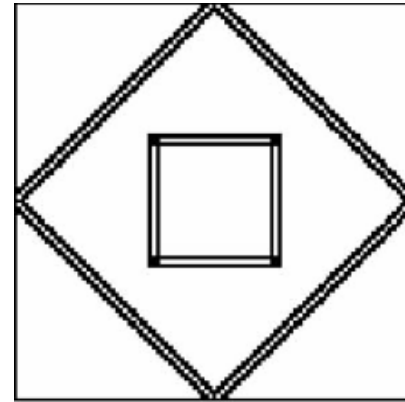
- However, both Canny and Sobel methods obtain zero intensity if a line of one – pixel width passes through the mask.
- Resulting in double edges as shown below



True Image



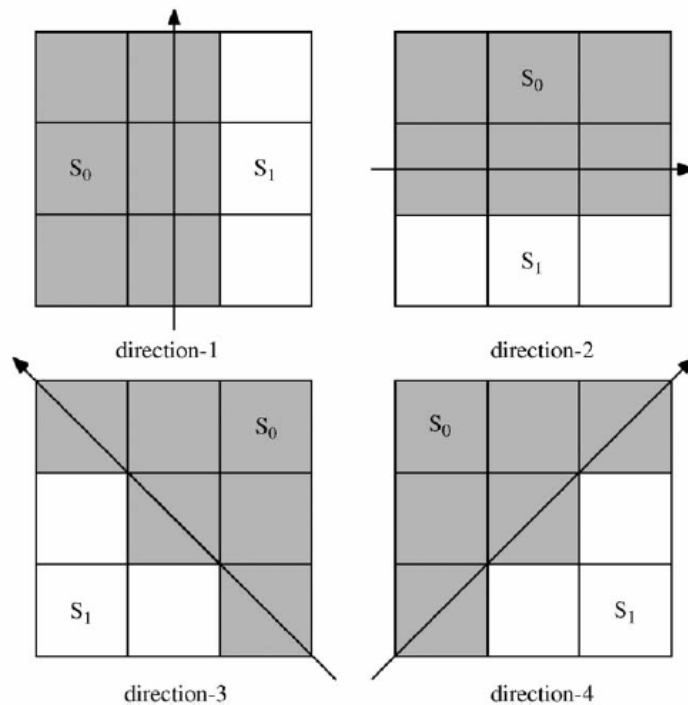
Sobel method

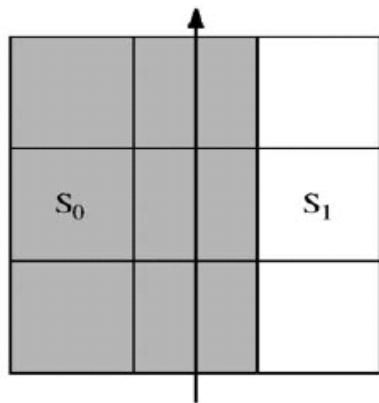


Canny method

Objective function method

- Algorithm follows an outline close to Canny
- Edge direction is decided by computing certain cost function





$$f_i = (L - 1) \frac{N_f}{D_f},$$

$$N_f = \min \left(1, \frac{|m_0 - m_1|}{w_1} \right),$$

$$D_f = 1 + \frac{1}{15} \sum_{\substack{p_m, p_n \in S_0 \\ m > n \\ m \neq n}} \min \left(1, \frac{|p_m - p_n|}{w_2} \right) + \frac{1}{3} \sum_{\substack{p_m, p_n \in S_1 \\ m > n \\ m \neq n}} \min \left(1, \frac{|p_m - p_n|}{w_2} \right)$$

$$E(x, y) = \max(f_1, f_2, f_3, f_4),$$

$$D(x, y) = \text{Arg}(\max(f_1, f_2, f_3, f_4)).$$

- Non Maxima Suppression

(1) $D(x, y) = 1$ and $E(x, y) > E(x, y - 1)$ and $E(x, y) > E(x, y + 1)$.

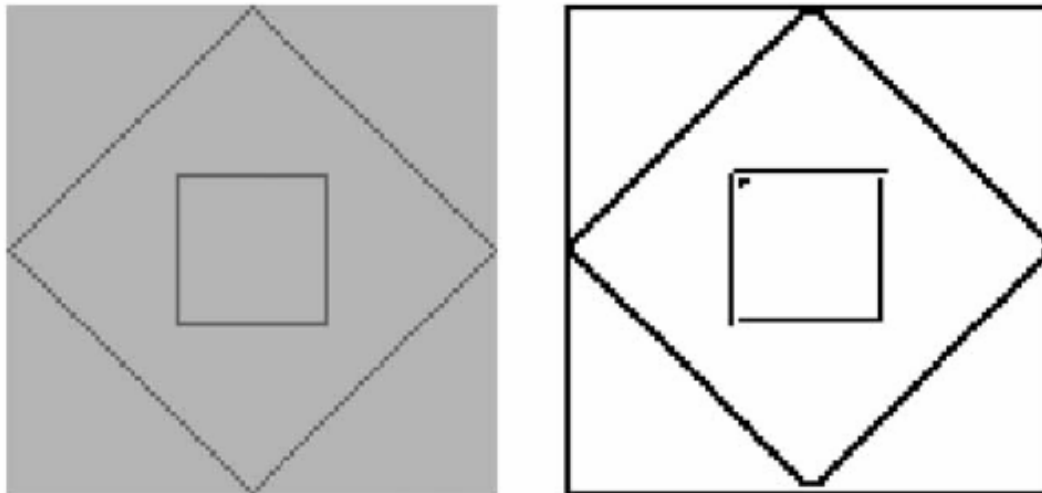
(2) $D(x, y) = 2$ and $E(x, y) > E(x - 1, y)$ and $E(x, y) > E(x + 1, y)$.

(3) $D(x, y) = 3$ and $E(x, y) > E(x - 1, y + 1)$ and $E(x, y) > E(x + 1, y - 1)$.

(4) $D(x, y) = 4$ and $E(x, y) > E(x - 1, y - 1)$ and $E(x, y) > E(x + 1, y + 1)$.

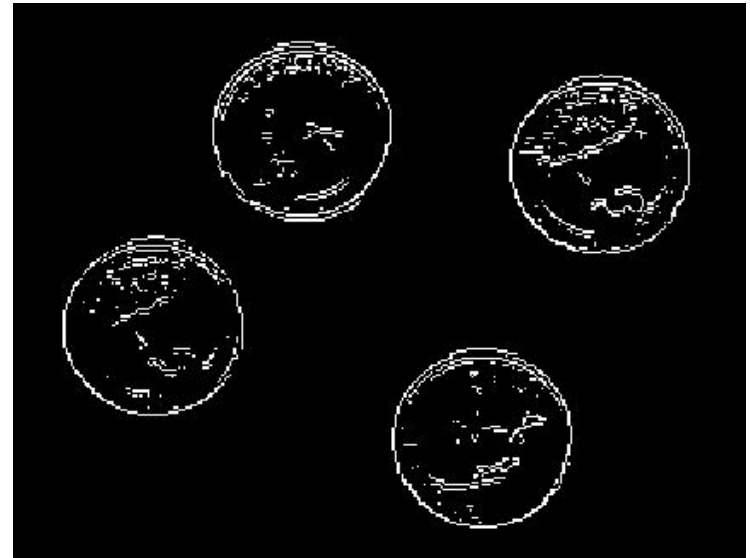
Pseudo Code

- Step 1 : Compute 4 objective function values
- Step 2 : Generate Edge map and Direction map
- Step 3 : Apply NMS to the edge and direction maps and extract the edge points in the image.





Original Image

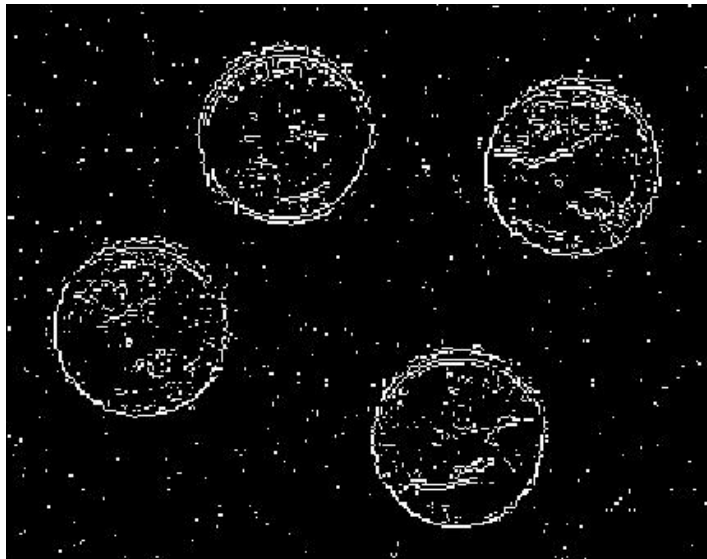


Edge Detected Image

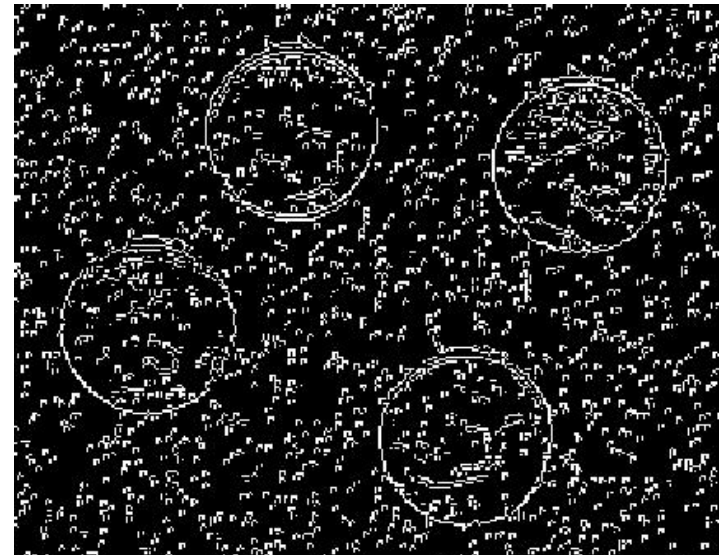
Drawbacks!

- w_0 and w_1 are constants determined from empirical results, not specific to an Image – leads to problems
- The algorithm outperforms Canny and Sobel edge detectors only in an ideal case.
- Fails to work efficiently in presence of Noise

Performance of the algorithm in noise



Gaussian Noise

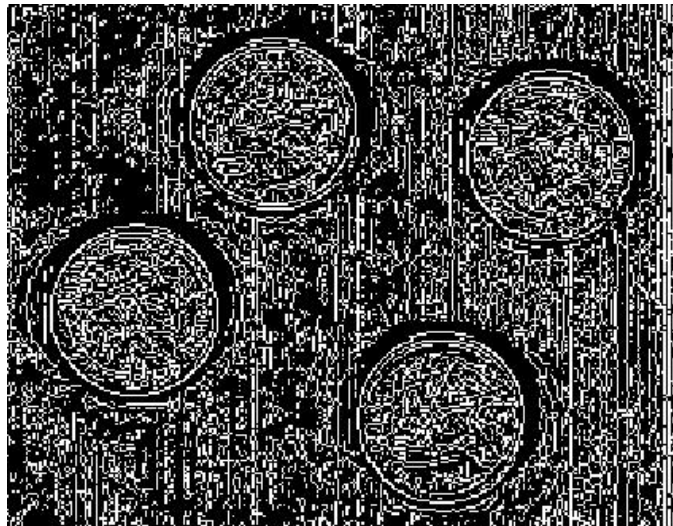


Salt and Pepper Noise

Proposed Improvements

A New cost function

- Original cost function depends on empirical estimates w_0, w_1
- Bad w_0, w_1 estimates will lead to poor edge detection
- We propose a more natural measure for inter-set and intra-set distances



Remodeling the cost function

- Cost function $f = N_f/D_f$

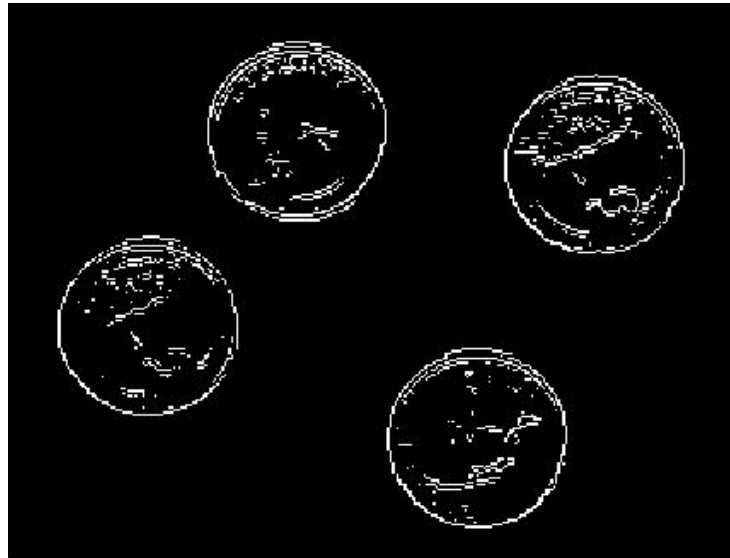
$$N_f = \sum_S N_S (\mu_S - \bar{x}_S)^2$$

$$D_f = 1 + \sum_S \sum_{i \in S} (x_i - \mu_S)^2$$

$$\mu_S = \frac{1}{N_S} \sum_{i \in S} x_i$$

$$\bar{x} = \frac{1}{N} \sum_i x_i$$

These distance measures have been borrowed from theory of LDA



Edge detected image using modified objective function

Objective function for edge detection in noise

- The drawback can be overcome by coming up with a clever choice of the objective function such that the existing framework can be used even in presence of noise
- We consider two different cases

Case 1 : Image in presence of AWGN

Case 2 : Image in presence of Salt and Pepper noise

Gaussian Noise

- How can I remove Gaussian noise?

Averaging!

- But, we don't want to blur the edges

Adaptive averaging – concept behind adaptive Wiener filtering

$$x^1(i, j) = \begin{cases} \mu(i, j) & \text{if } |\mu(i, j) - x^0(i, j)| < a \\ x^0(i, j) & \text{if } |\mu(i, j) - x^0(i, j)| > a \end{cases}$$

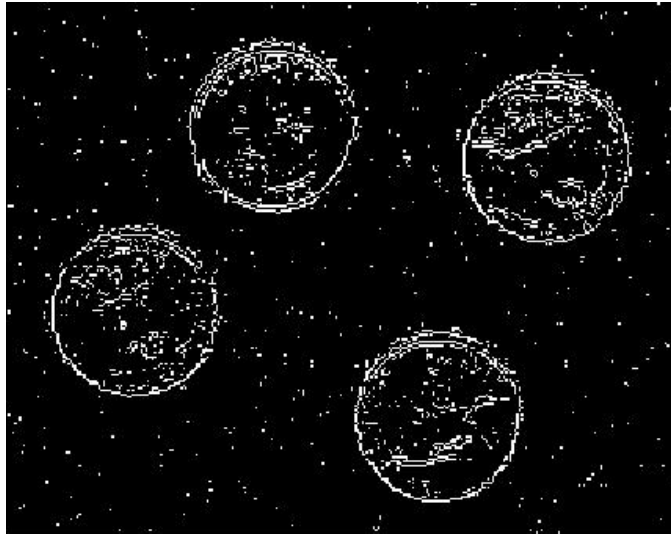
- Incorporate this into the objective function

Modified cost function for Gaussian Noise

- Cost function $f = N_f/D_f$

$$N_f = \begin{cases} \sum_S N_S (\mu_S - \bar{x}_S)^2 & \text{if } |\mu_S - \bar{x}_S| > a \\ 0 & \text{if } |\mu_S - \bar{x}_S| < a \end{cases}$$

$$Df = \begin{cases} 1 + \sum_S \sum_{i \in S} (x_i - \mu_S)^2 & \text{if } |x_i - \mu_S| > a \\ 1 & \text{else} \end{cases}$$



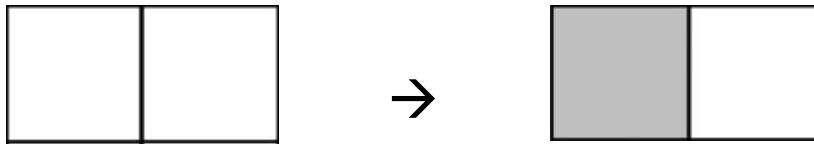
Original Algorithm



Proposed Algorithm

Salt and Pepper Noise

- Unlike Gaussian noise, Impulse/Salt and Pepper noise corrupts the image with very high values.
- Resulting in many false edges! - Why?



Normal edge detection = $f(i,j+1) - f(i,j)$

- Solution :
Proposed edge detection = $\log(\text{int}(\alpha f(i,j+1))) - \log(\text{int}(\alpha f(i,j)))$
where 'int' stands for greatest integer

Remodeling the cost function

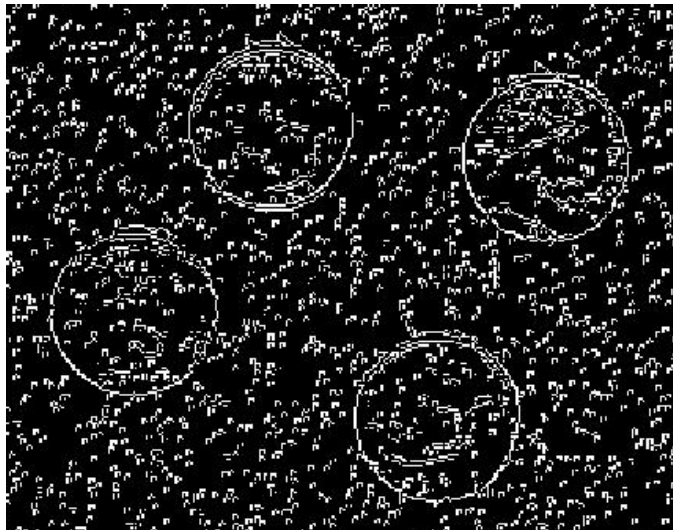
- Cost function $f = N_f/D_f$

$$\text{where } N_f = \sum_S N_S (\log \mu_S - \log \bar{x}_S)^2$$

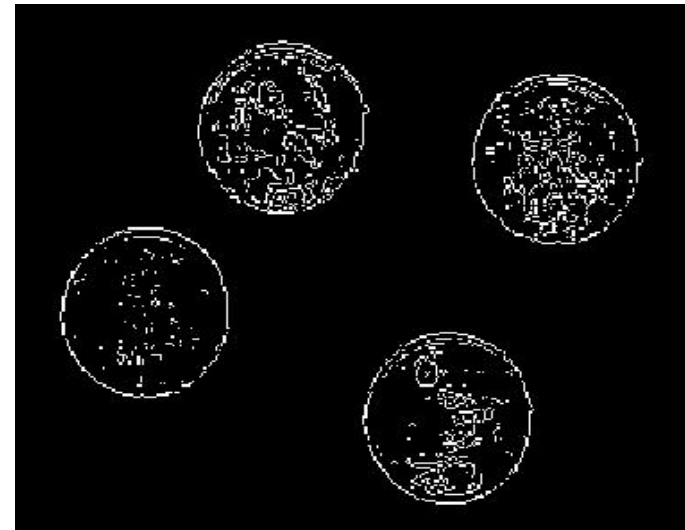
$$D_f = 1 + \sum_S \sum_{i \in S} (\log x_i - \log \mu_S)^2$$

$$\mu_S = \frac{1}{N_S} \sum_{i \in S} x_i$$

$$\bar{x} = \frac{1}{N} \sum_i x_i$$



Original Algorithm

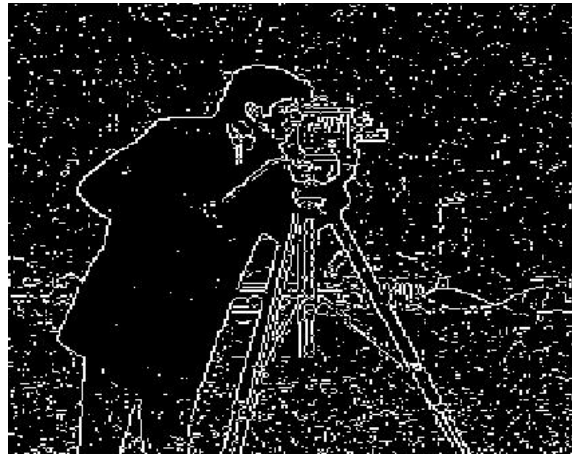


Proposed Algorithm

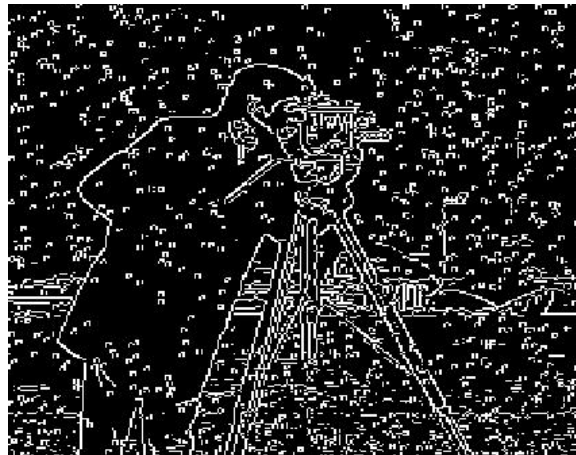
Original Algorithm

Proposed Algorithm

Gaussian Noise



Salt and Pepper noise



Conclusions

- Objective function based edge detection method works better than traditional edge detection methods.
 - Not robust to noise
 - Performance in noise can be improved by simple changes to objective function
 - Versatile
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Another application ?

- Edges in color images can be classified as object edges, shadow edges, reflectance edges, occlusion edges etc.
 - There is a lot of recent work concerning classification of edges
 - Our aim – use different objective functions to distinguish and classify different edges....
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References

- [1] 'A novel edge detection method based on maximising objective functions' , Kang, Wang, Journal of Pattern Recognition, 2006
 - [2] 'A new edge detection method in Image Processing', Zhang, Zhao, Su, Proc of ISCIT, 2005
 - [3] ' Color edge detection in presence of Gaussian noise using non linear pre filtering', Russo, Lazzari, IEEE Transactions on Inst. and Measurements, 2005.
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Thank You

Questions?
