

# Analisi di Immagini e Video (Computer Vision)

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# Crediti

- Slides adattate da vari corsi
  - Analisi di Immagini (F. Angiulli) – Unical
  - Intro to Computer Vision (J. Tompkin) – CS Brown Edu
  - Computer Vision (I. Gkioulekas), CS CMU Edu

# Recap: Gradient-based filtering

1. Scegli la derivata

$$\mathbf{S}_x = \begin{array}{|c|c|c|} \hline 1 & 0 & -1 \\ \hline 2 & 0 & -2 \\ \hline 1 & 0 & -1 \\ \hline \end{array}$$

$$\mathbf{S}_y = \begin{array}{|c|c|c|} \hline 1 & 2 & 1 \\ \hline 0 & 0 & 0 \\ \hline -1 & -2 & -1 \\ \hline \end{array}$$

2. Convolvi con l'immagine

$$\frac{\partial f}{\partial x} = \mathbf{S}_x * f$$

$$\frac{\partial f}{\partial y} = \mathbf{S}_y * f$$

3. Calcola direzione e ampiezza del gradiente.

$$\nabla f = \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$

gradient

$$\theta = \tan^{-1} \left( \frac{\partial f}{\partial y} / \frac{\partial f}{\partial x} \right)$$

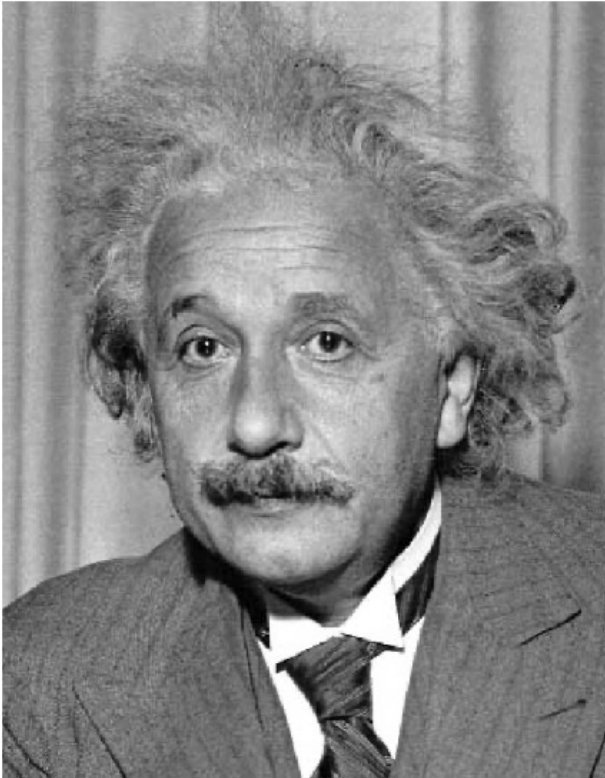
direction

$$\|\nabla f\| = \sqrt{\left( \frac{\partial f}{\partial x} \right)^2 + \left( \frac{\partial f}{\partial y} \right)^2}$$

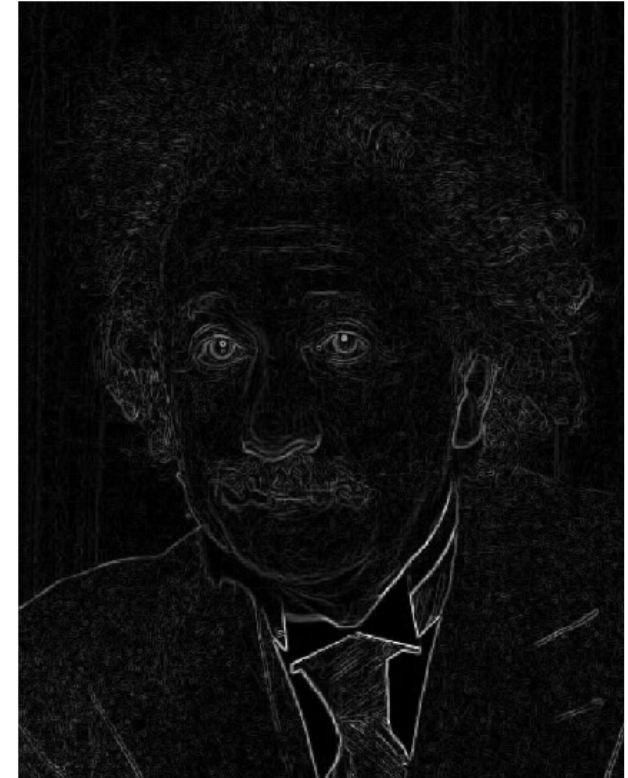
amplitude

# Gradient-Based Filtering

originale



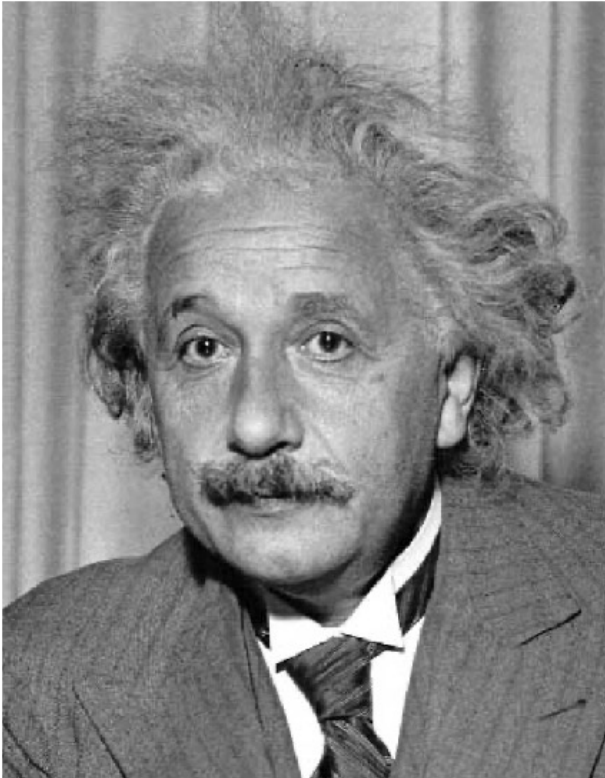
Ampiezza



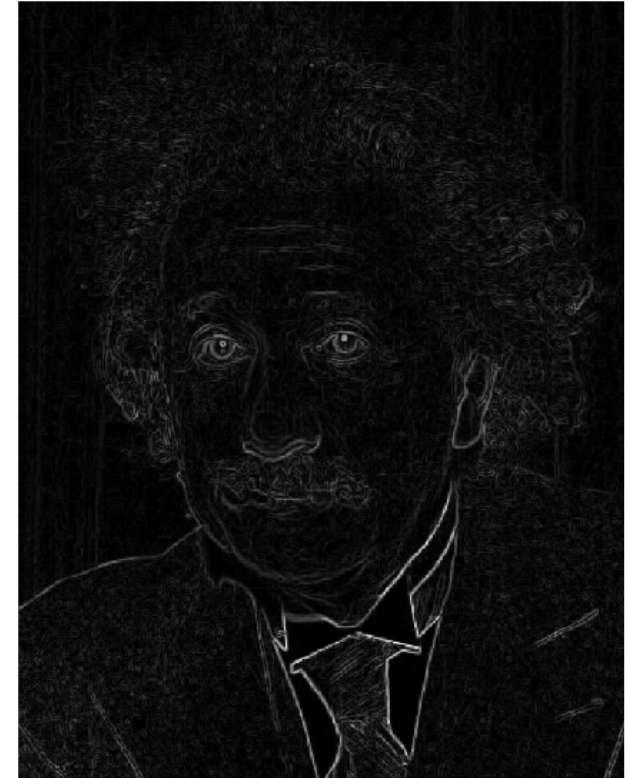
- Cosa c'è che non va?

# Gradient-Based Filtering

originale



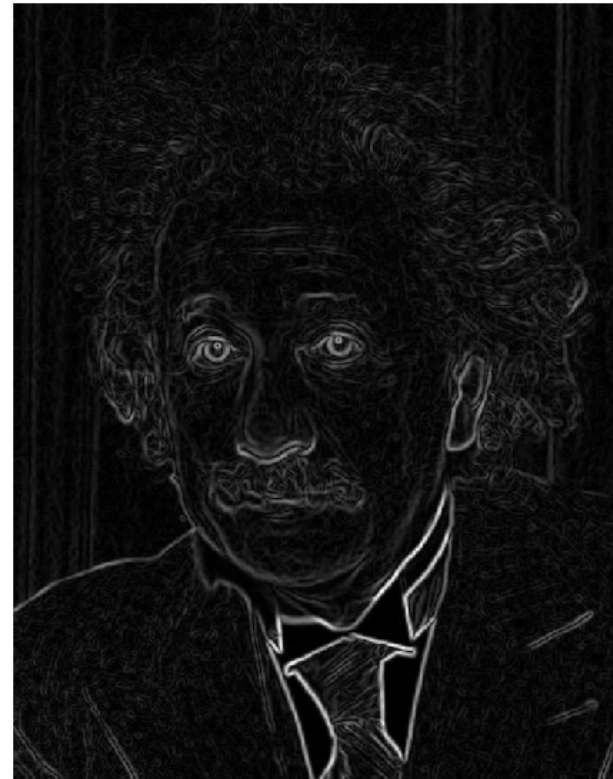
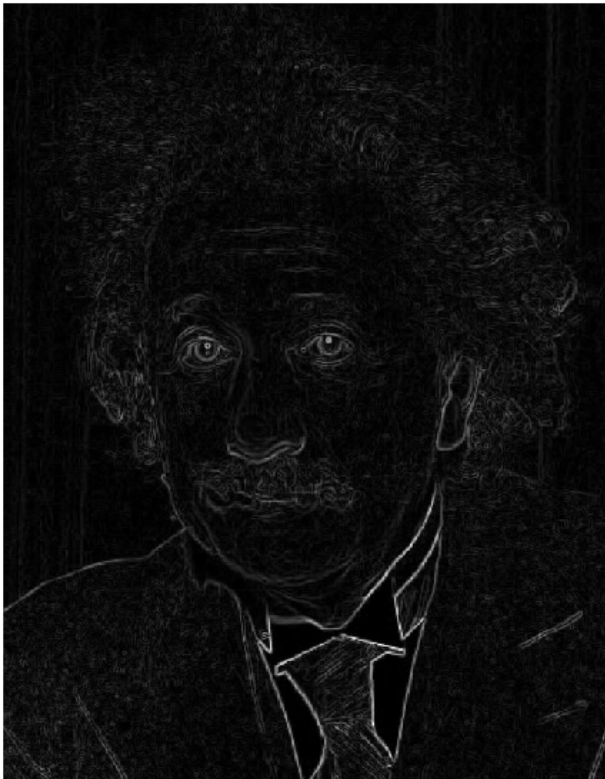
Ampiezza



- Cosa c'è che non va?
  - Troppi artefatti

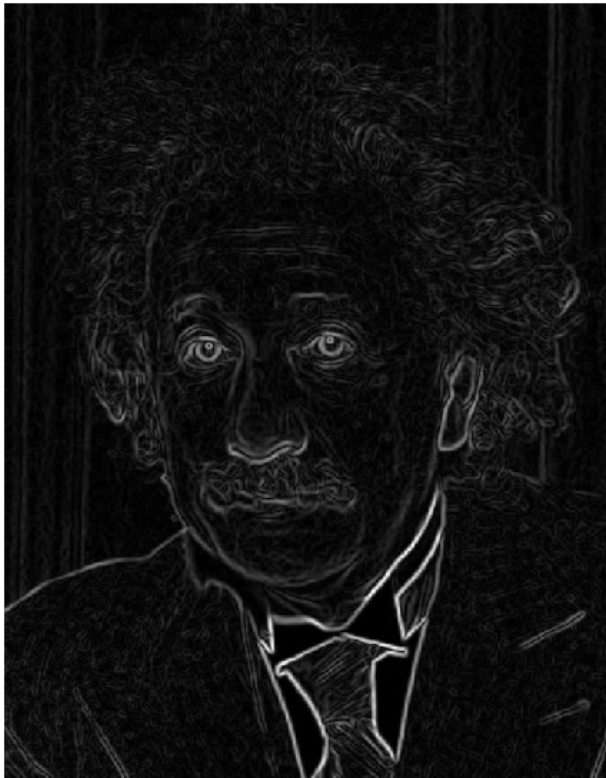
# Gradient-Based Filtering

- Noise
  - Lo trattiamo con il filtro gaussiano



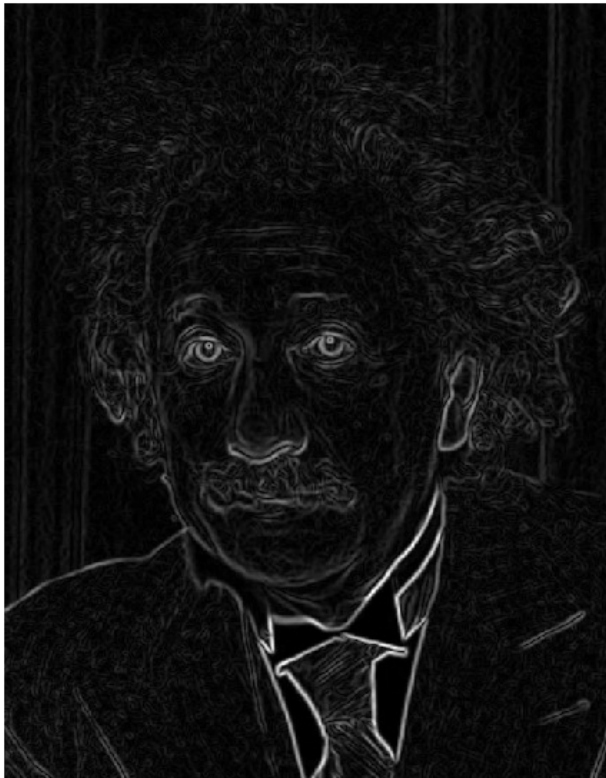
# Gradient-Based Filtering

- Troppi toni di grigio
  - Thresholding



# Gradient-Based Filtering

- Diversi spessori
  - ?



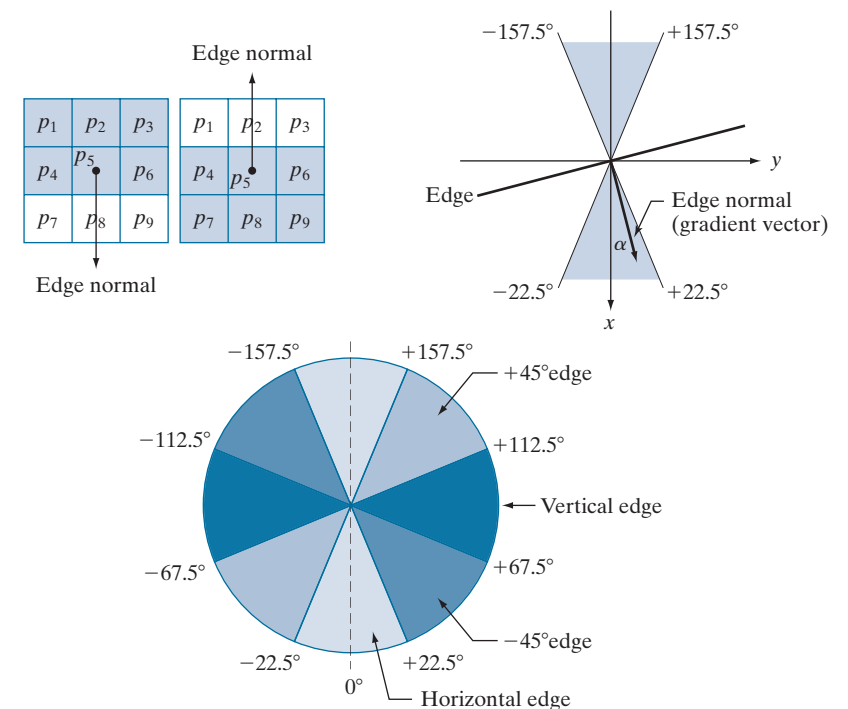
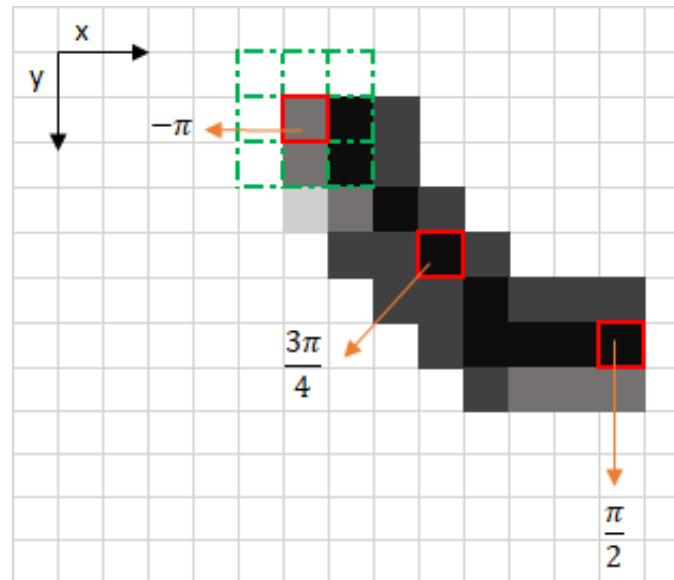


# Canny Edge Detector

- Algoritmo semplice
  - Estensione del gradient-based filtering
  - Spessori uniformi
    - Non-maximal suppression
  - Rimozione di artefatti
    - Double thresholding, hysteresis

# Non maximal suppression

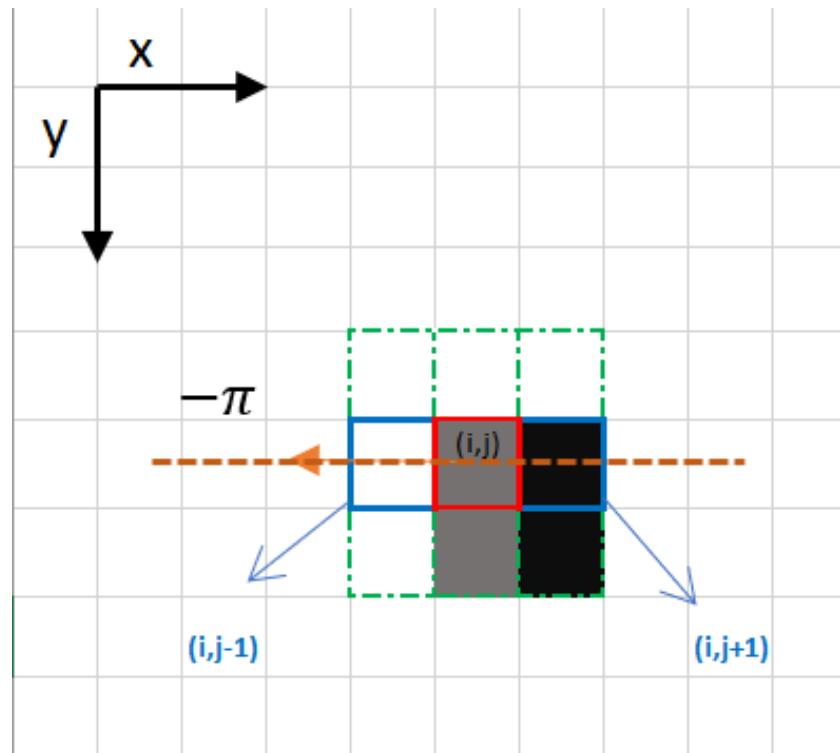
- Basata sulla direzione del gradiente  $\theta = \tan^{-1} \left( \frac{\partial f}{\partial y} / \frac{\partial f}{\partial x} \right)$



- Il gradiente punta a valori alti di intensità

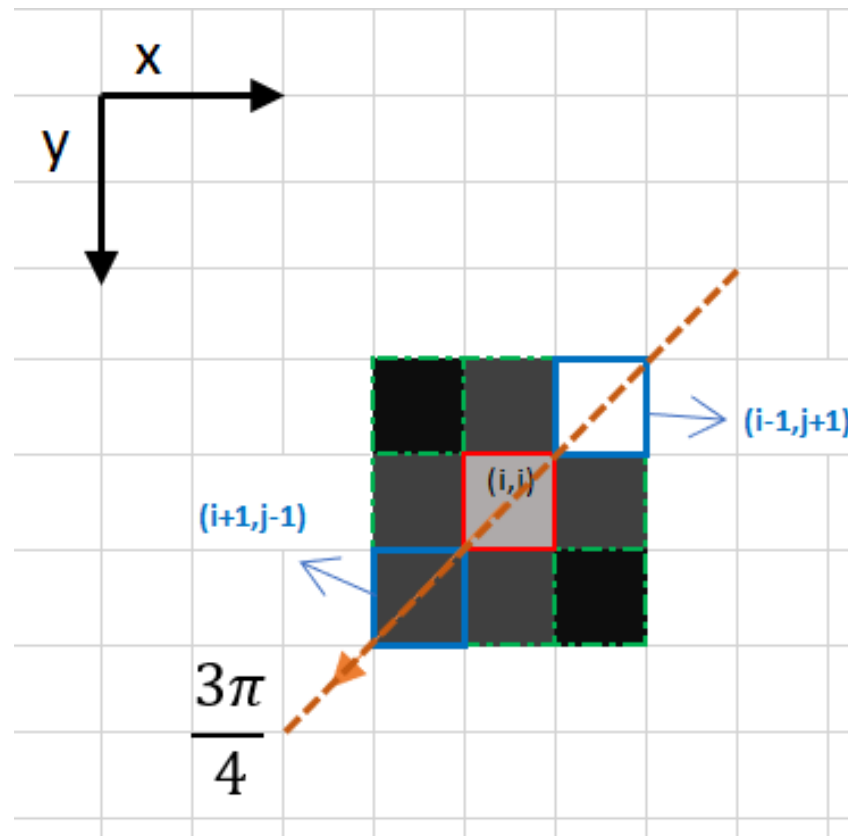
# Non maximal suppression

- Sopprimiamo il pixel se lungo la direzione del gradiente i pixel adiacenti hanno un valore più alto

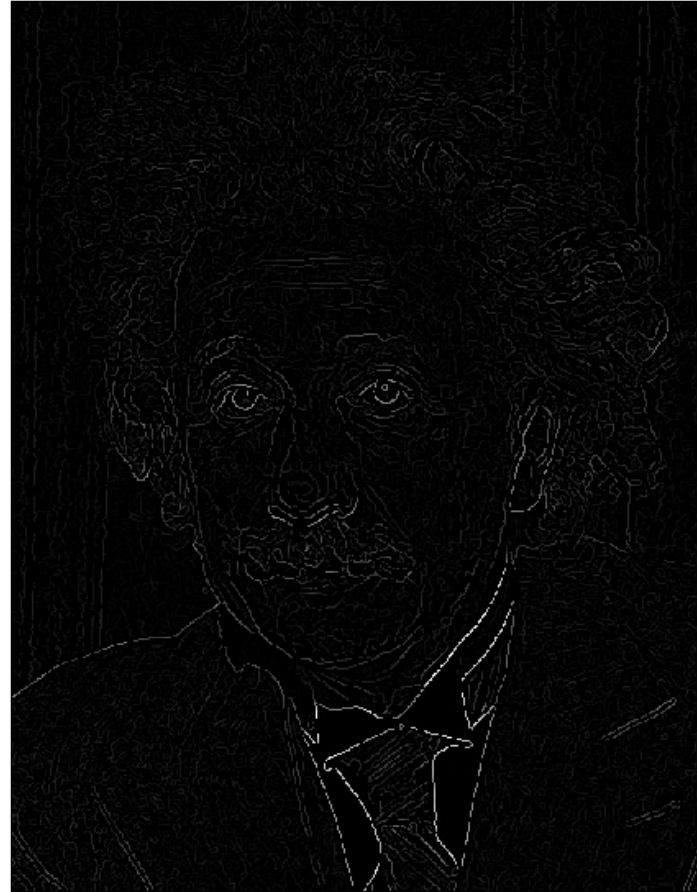
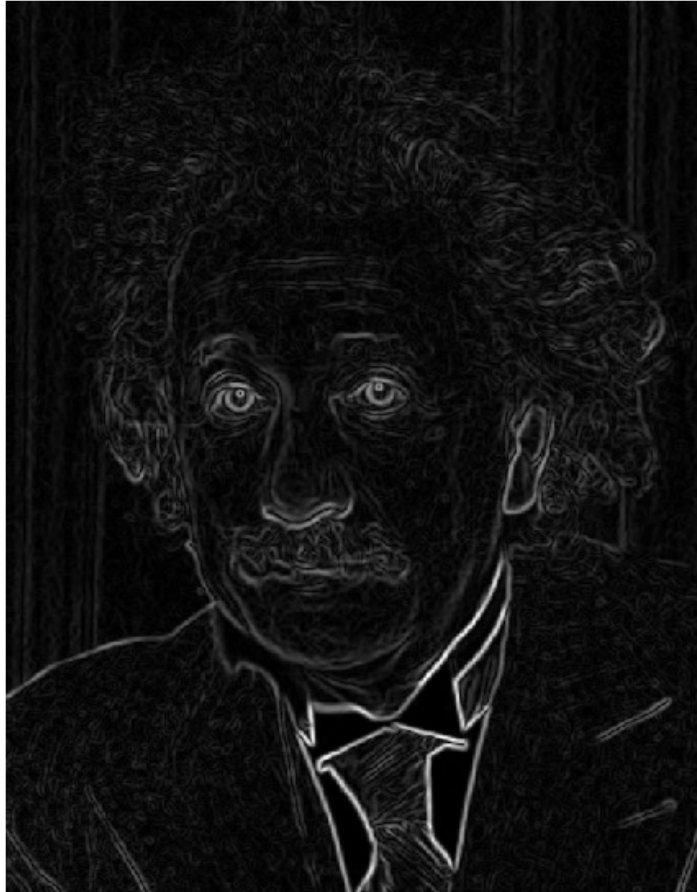


# Non maximal suppression

- Sopprimiamo il pixel se lungo la direzione del gradiente i pixel adiacenti hanno un valore più alto



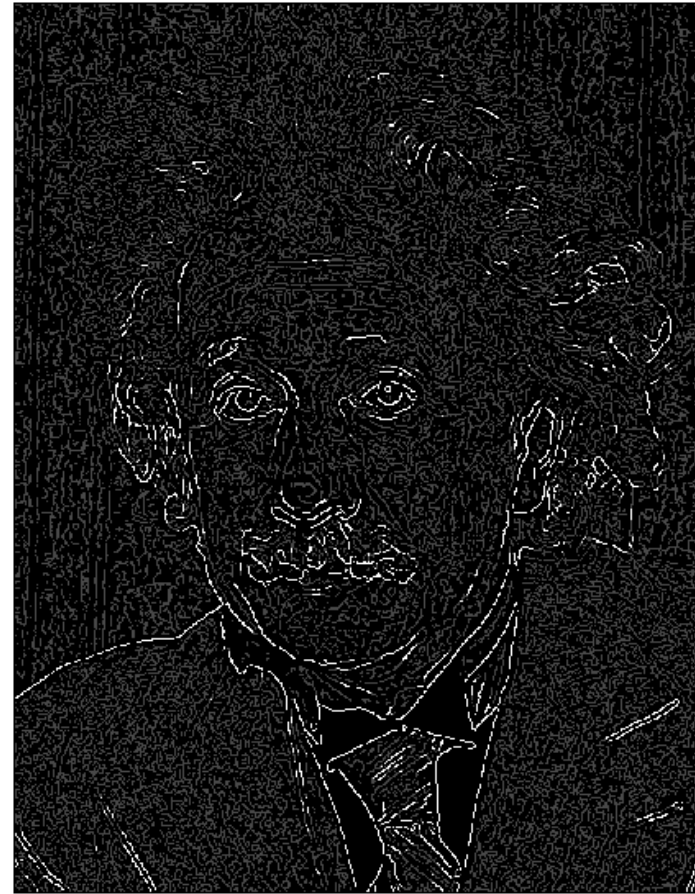
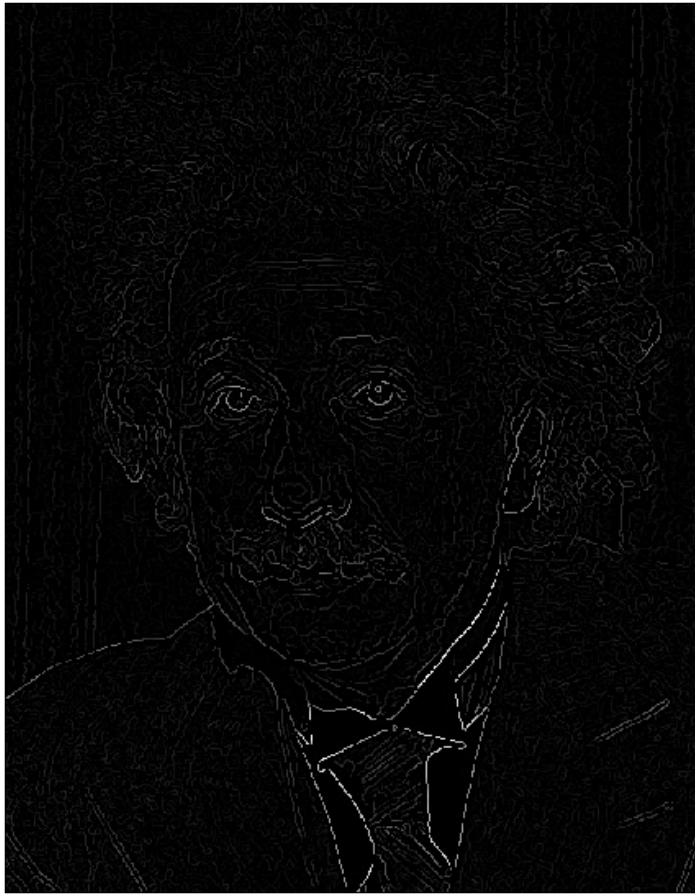
# Non-Maximal Suppression



# Level-wise thresholding, hysteresis

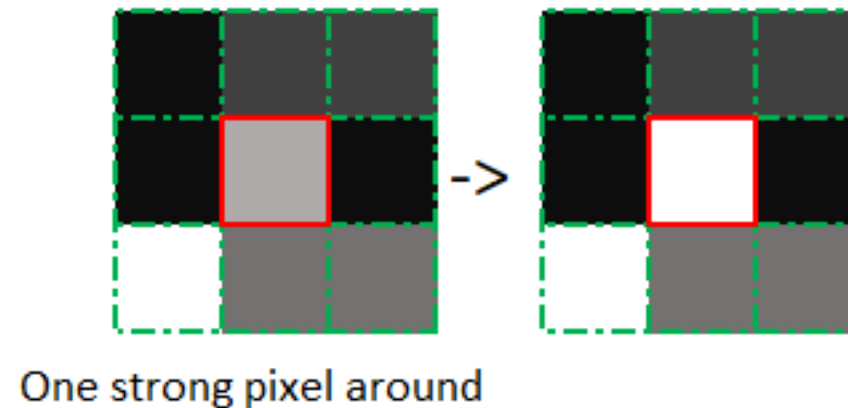
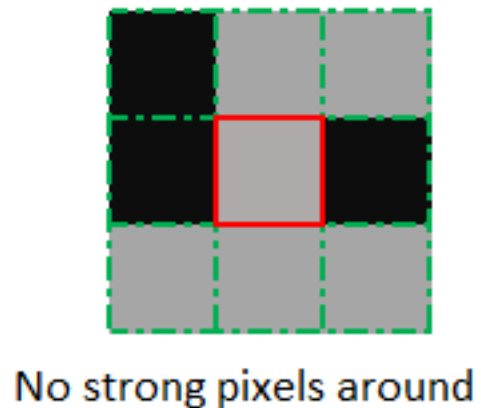
- Strong/Weak/irrelevant pixels
  - Strong pixels: intensità alta (contribuiscono sicuramente ai bordi)
  - Weak pixels: intensità non alta, ma neanche bassa
    - Li teniamo da parte
  - Irrelevant pixels: intensità bassa, da rimuovere
- Usiamo due soglie
  - High threshold per identificare strong pixels
  - Low threshold per identificare irrelevant pixels
- Tutti i pixel nel mezzo delle due soglie sono weak e verranno gestiti dal meccanismo dell'isteresi

# Level-wise thresholding



# Hysteresis

- «Attrazione gravitazionale»
  - Se un weak Pixel ha uno strong pixel nel vicinato, diventa anch'esso uno strong pixel, altrimenti diventa irrelevant e viene soppresso





# Hysteresis

