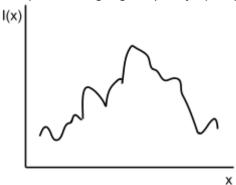
Edge Detection

- What can cause change in intensity of light?
 - Reflectance change
 - Shadows: relatively abrupt change in illumination
 - Occlusions
 - Convex + Concave surface edges
 - Specular highlights (shiny spots)



- How to find changes in intensity?
 - Find changes in intensity by measuring slopes, which is the derivative of the image.

•
$$f'(x) = \lim_{h \to 1} \frac{f(x+h) - f(x)}{h} = > \frac{f(x+1) - f(x)}{1}$$

• Convolution mask [1 | -1] OR [-1 | 1] can be used to find $f'(x)$

- horizontally
- In two dimensional image f(x, y), use the mask both horizontally and vertically

$$\frac{\partial f(x,y)}{\partial x} = \lim_{h \to 1} \frac{f(x+h,y) - f(x,y)}{h}$$

$$\frac{\partial f(x,y)}{\partial y} = \lim_{h \to 1} \frac{f(x,y+h) - f(x,y)}{h}$$

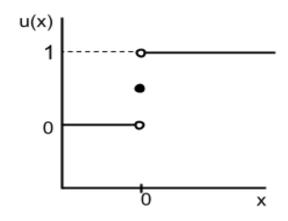
$$\bullet \quad \frac{\partial f(x,y)}{\partial y} = \lim_{h \to 1} \frac{f(x,y+h) - f(x,y)}{h}$$

Combining the above two derivatives together, the Gradient is given by

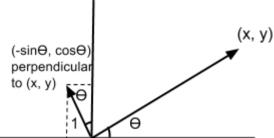
$$\nabla f = \left(\frac{\partial f(x,y)}{\partial x}, \frac{\partial f(x,y)}{\partial y}\right)$$

- Magnitude of the gradient is the 'slope' we need. To find edges we find where the magnitude of the gradient is big.
- Problem
 - Derivative is very 'local' because we take only two neighboring pixels and calculate the gradient, we may end up finding many wiggles with big gradient too.
- Consider theoretical case of an image with just two intensities
 - Define step function

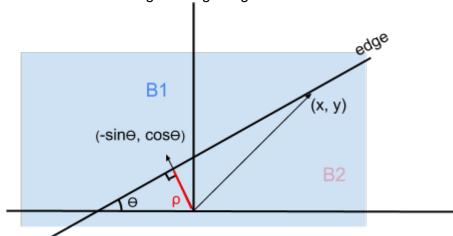
$$u(x) = \begin{cases} 1 & x > 0 \\ \frac{1}{2} & x = 0 \\ 0 & x < 0 \end{cases}$$



■ A line through origin: $(x, y) * (-\sin\theta, \cos\theta) = 0$ → dot product of perpendicular vectors is zero



A line which doesn't go through origin



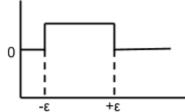
- $E(x,y) = B_1 + [B_2 B_1]u(x * sin\theta y * cos\theta + \rho)$ $u = 0 \Rightarrow E(x,y) = B_1$ $u = 1 \Rightarrow E(x,y) = B_2$ $\frac{\partial E}{\partial x} = (B_2 B_1)sin\theta * u'(xsin\theta ycos\theta + \rho)$ $\frac{\partial E}{\partial y} = (B_2 B_1)(-cos\theta) * u'(xsin\theta ycos\theta + \rho)$
- Consider the step function as the limit of the sequence $\{u_{\epsilon}(x)\}$ where

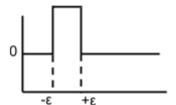
$$u_{\epsilon}(x) \begin{cases} 1 & x > \epsilon \\ \frac{1}{2}(1 + x/\epsilon) & |x| < \epsilon \\ 1 & x < -\epsilon \end{cases}$$

- Taking the derivative of each $\{u_{\epsilon}(x)\}$ gives a sequence the limit of which is the Dirac Delta. We use it as a point source, written as δ
- As $\epsilon \to 0$

$$u'_{\epsilon}(x) = \frac{du_{\epsilon}(x)}{dx} = \begin{cases} \frac{1}{2\epsilon} & |x| \le \epsilon \\ 0 & |x| > \epsilon \end{cases}$$

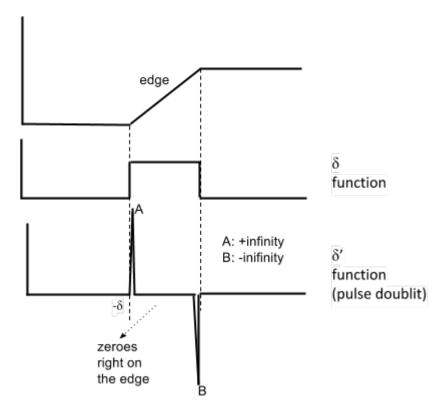
What does $1/2\epsilon$ looks like?





- When ϵ gets smaller, the central part gets taller
- Area under the curve = $2\epsilon * (1/2\epsilon) = 1$
- $\begin{array}{ll} \bullet & \frac{\partial E}{\partial x} = sin\theta (B_2 B_1) \delta (xsin\theta ycos\theta + \rho) \\ \bullet & \frac{\partial E}{\partial y} = -cos\theta (B_2 B_1) \delta (xsin\theta ycos\theta + \rho) \end{array}$
- δ has a 'sifting' property: $\int_{-\infty}^{+\infty} \delta(x)h(x)dx = h(0)$
- $\nabla E = (\frac{\partial E}{\partial x}, \frac{\partial E}{\partial y}) = (B_2 B_1)(sin\theta cos\theta) \rightarrow \text{perpendicular to the}$ edge
- Problem
 - The same scene with difference in brightness may result in different gradients. This can be eliminated by threshold, but using a threshold can neglect some local maximums of gradient.
- Find local maxima or minima in I'(x) where I''(x) = 0

 - $\frac{\partial^{2} E}{\partial x^{2}} = \sin^{2}\theta (B_{2} B_{1})\delta'(x\sin\theta y\cos\theta + \rho)$ $\frac{\partial^{2} E}{\partial y^{2}} = \cos^{2}\theta (B_{2} B_{1})\delta'(x\sin\theta y\cos\theta + \rho)$
 - Laplacian: $\nabla^2 E = \frac{\partial^2 E}{\partial x^2} + \frac{\partial^2 E}{\partial y^2} = \delta'(B_2 B_1)(xsin\theta ycos\theta + \rho)$
 - To find local maxima/minima we need to locate where $\nabla^2 E$ is 0 by finding where δ' is 0



- We look for the zero crossings: the segment between A and B in the above graph is the edge.
- How to find Laplacian in discrete case

$$\bullet \quad \frac{\partial E}{\partial x} = \lim_{h \to 1} \frac{E(x+h) - E(x)}{h} = E(x+1) - E(x) \Rightarrow \text{use [-1]+1] mask}$$

■
$$\frac{\partial E}{\partial x} = \lim_{h \to 1} \frac{E(x+h) - E(x)}{h} = E(x+1) - E(x)$$
 use [-1|+1] mask
■ $\frac{\partial^2 E}{\partial x^2} = \lim_{h \to 1} \frac{\frac{dE(x+h)}{h} - \frac{dE(x)}{dx}}{h} = E(x+2) - 2E(x+1) + E(x)$
 \Rightarrow use [1|-2|1] mask

Combining the mask horizontally and vertically we get

Combining			
0	1	0	
1	-4	1	
0	1	0	

- Strategy in edge detection: 1. Smooth the image 2. Take the Laplacian
 - Smoothing (blurring) the image using weighted average
 - Gaussian in 2D

$$G(x,y) = -\frac{1}{2\pi\sigma^2} e^{-\frac{x^2 + y^2}{2\sigma^2}}$$

$$G(r) = -\frac{1}{2\pi\sigma^2} e^{-\frac{r^2}{2\sigma^2}} \text{ where } r = x^2 + y^2$$

- Input image → Output weighted average of each pixel
- σ controls the output \rightarrow the bigger σ is, the bigger the mask is, the more blurring the output is
- ❖ $\nabla^*(G^*E) = (\nabla^*G)^*E$ → Apply the Laplacian of Gaussian to different images