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Problem 3: Proof that convolutions are in the continuous case associative.

Proof:

Definition of Convolution:

$$(f * g)(x) = \int_0^x f(x-u)g(u) du = \int_0^x f(u)g(x-u) du.$$

Let another convolution exists such that

$$((f * g) * h)(x) = \int_0^x ((f * g)(\delta)) h(x-\delta) d\delta$$

$$= \int_0^x \left( \int_0^\delta f(u)g(\delta-u) du \right) h(x-\delta) d\delta$$

$$= \int_0^x \int_0^\delta f(u)g(\delta-u) h(x-\delta) du d\delta$$

$$= \int_0^x \int_u^x f(u)g(\delta-u) h(x-\delta) d\delta du$$

$$= \int_0^x \int_0^{x-u} f(u)g(\delta-u) h(x-\delta) d\delta du$$

$$= \int_0^x f(u) \left( \int_0^{x-u} g(\delta-u) h(x-\delta) d\delta \right) du$$

$$= \int_0^x f(u) ((g * h)(x-u)) du$$

$$= f * (g * h)(x) \quad \underline{\text{proved}}$$

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Problem 6: Proof that convolution two times with a Gaussian kernel with standard deviation  $\sigma$  is the same as convolution once with standard deviation  $\sqrt{2}\sigma$ .

Proof:

Firstly, We have the following Gaussian Kernel with std. deviation  $\sigma$ .

$$G_{\sigma} = \frac{1}{2\pi\sigma^2} e^{-\frac{(x^2+y^2)}{2\sigma^2}} = \underbrace{\left( \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}} \right)}_{\text{1D Gaussian}} \left( \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{y^2}{2\sigma^2}} \right)$$

Let the original image is  $I$

Applying convolution to image  $I$  with Gaussian kernel  $G_{\sigma}$ , we have.

$$I * G_{\sigma} \quad \text{--- (1)}$$

Applying convolution again to eq<sup>n</sup> (1) we have

$(I * G_{\sigma}) * G_{\sigma}$ . which according to associative property of convolution is as follows:

$$(I * G_{\sigma}) * G_{\sigma} = I * (G_{\sigma} * G_{\sigma})$$

i.e; convolving pixel  $x$  twice resulted in:

$$\begin{aligned} I * \int_0^u \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-u)^2}{2\sigma^2}} \cdot \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{u^2}{2\sigma^2}} du & \quad g(x) = \int f(x-u)h(u)du \\ & = I * \frac{1}{2\pi\sigma^2} \int_0^u e^{-\frac{(x-u)^2}{2\sigma^2} + \frac{(-u^2)}{2\sigma^2}} du & = I * \frac{1}{2\pi\sigma^2} \int_0^u e^{-\frac{x^2 - u^2 - u^2}{2\sigma^2}} du \\ & = I * \frac{1}{2\pi\sigma^2} \int_0^u e^{-\frac{x^2 + 2xu - u^2 - u^2}{2\sigma^2}} du \\ & = I * \frac{1}{2\pi\sigma^2} \int_0^u e^{-\frac{x^2 - 2u^2 + 2xu}{2\sigma^2}} du \\ & = I * \frac{1}{2\pi\sigma^2} \cdot e^{-\frac{x^2}{2\sigma^2}} \cdot \int_0^u e^{\frac{2u(x-u)}{2\sigma^2}} du \\ & = I * \frac{1}{2\pi\sigma^2} \cdot e^{-\frac{x^2}{2\sigma^2}} \int_0^u e^{\frac{u(x-u)}{\sigma^2}} du \\ & = I * \frac{1}{2\pi\sigma^2} \cdot e^{-\frac{x^2}{4\sigma^2}} \\ & = I * \frac{1}{\sqrt{2\pi}(\sqrt{2}\sigma)^2} \cdot e^{-\frac{x^2}{2(\sqrt{2}\sigma)^2}} \\ & = I * G' \end{aligned}$$

In  $G'$  new value of <sup>std. deviation</sup>  $\sigma$  after convolving twice becomes  $\sqrt{2}\sigma$ .

$\therefore$  Convolution with Gaussian kernel with std. dev.  $\sigma$  two times is equal to convolution with Gaussian kernel with std. dev.  $\sqrt{2}\sigma$ .

proved