

## Image Blurring and the Convolution Theorem

Convolution theorem states that convolution in spatial domain is equivalent to pointwise multiplication of frequency domain

$$f * g = \mathcal{F}^{-1} \{ \mathcal{F}\{f\} \cdot \mathcal{F}\{g\} \}$$

Where

$f$  is the input image

$g$  is the convolution kernel

$\mathcal{F}$  denotes the Fourier transformation

$\mathcal{F}^{-1}$  denotes the Inverse Fourier Transformation

$*$  denotes convolution

$\cdot$  denotes element-wise multiplication

For a 2D image  $I(x, y)$  and kernel  $K(x, y)$

$$(I * K)(x, y) = \sum_{i=-a}^a \sum_{j=-b}^b I(x-i, y-j) \cdot K(i, j)$$

The equivalent operation in the frequency domain:

$$I_{\text{blurred}} = \mathcal{F}^{-1} \{ \mathcal{F}\{I\} \cdot \mathcal{F}\{K\} \}$$

Using a normalized Gaussian kernel:

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

For a  $15 \times 15$  kernel with  $\sigma = 2.5$ , the kernel is normalized so  $\sum G(x, y) = 1$