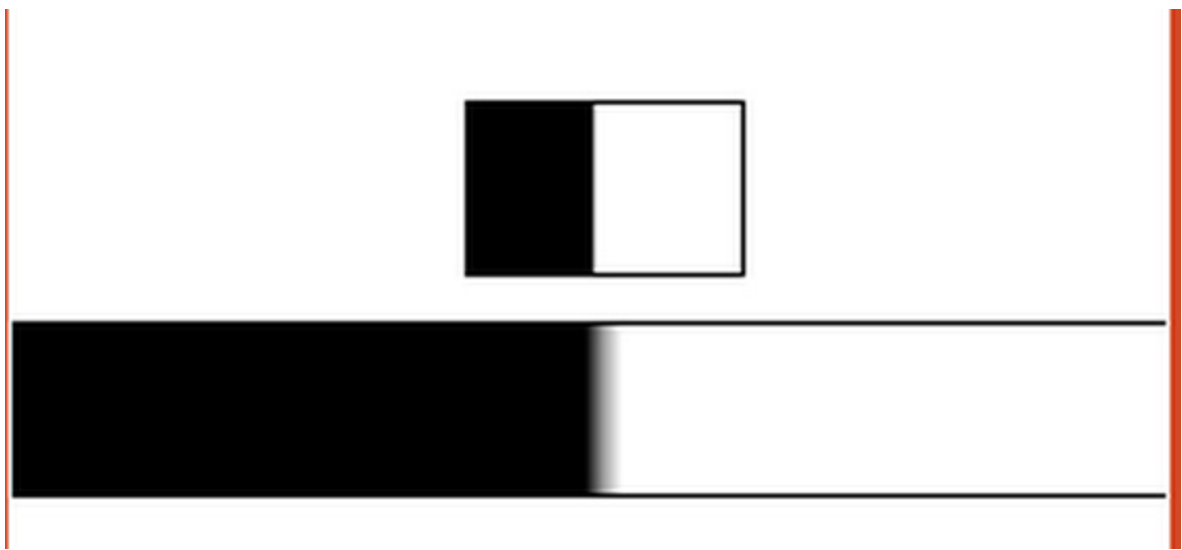
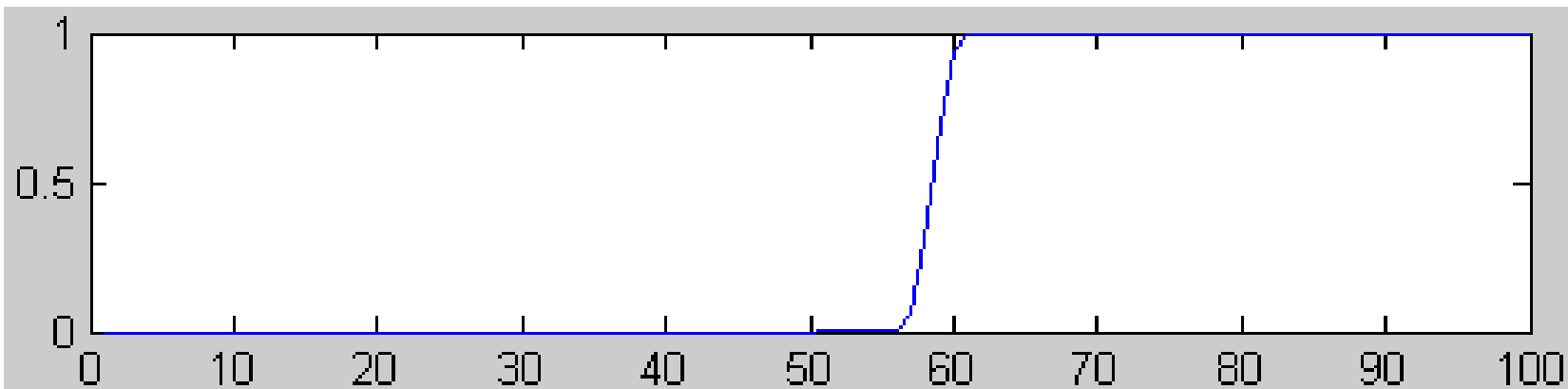


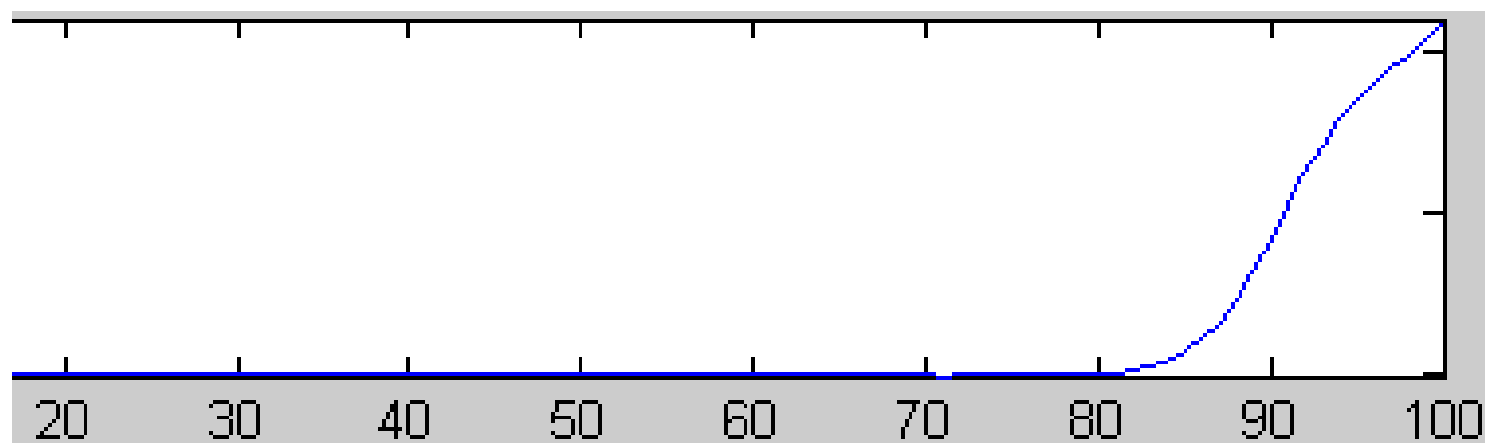
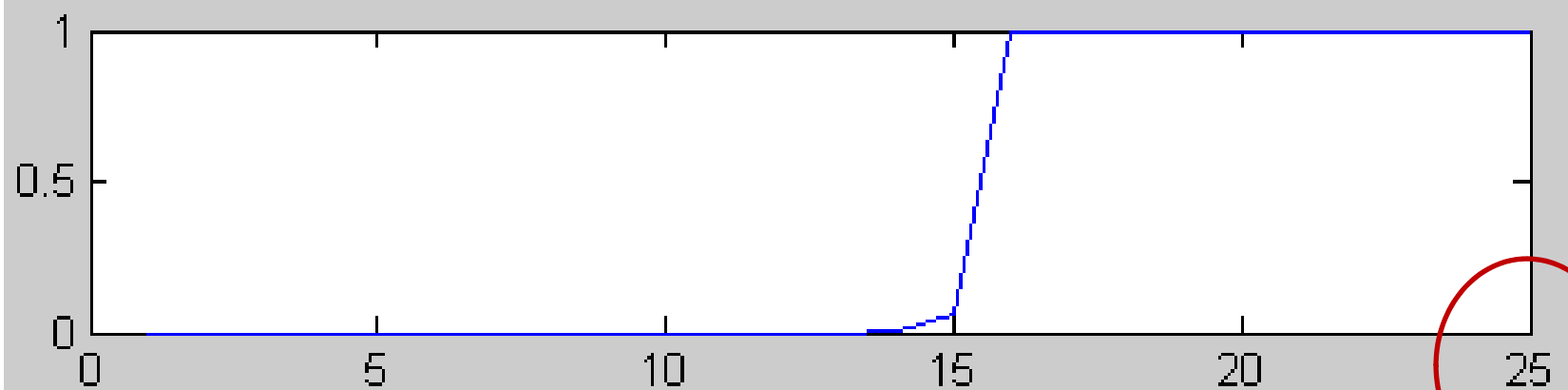
**The SUPER_RESOLUTION
EDGE RECONSTRUCTION
+ Restoration**



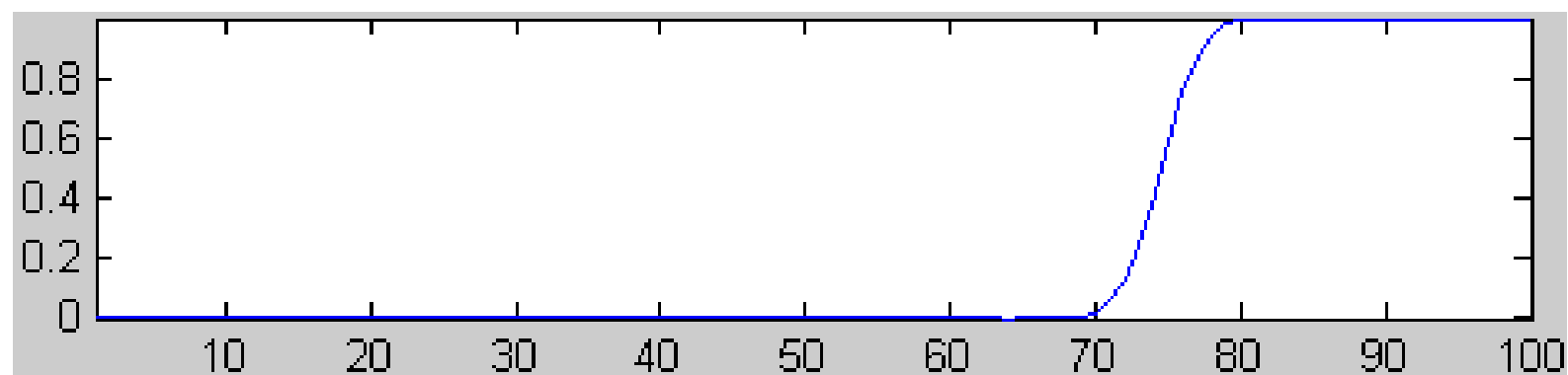
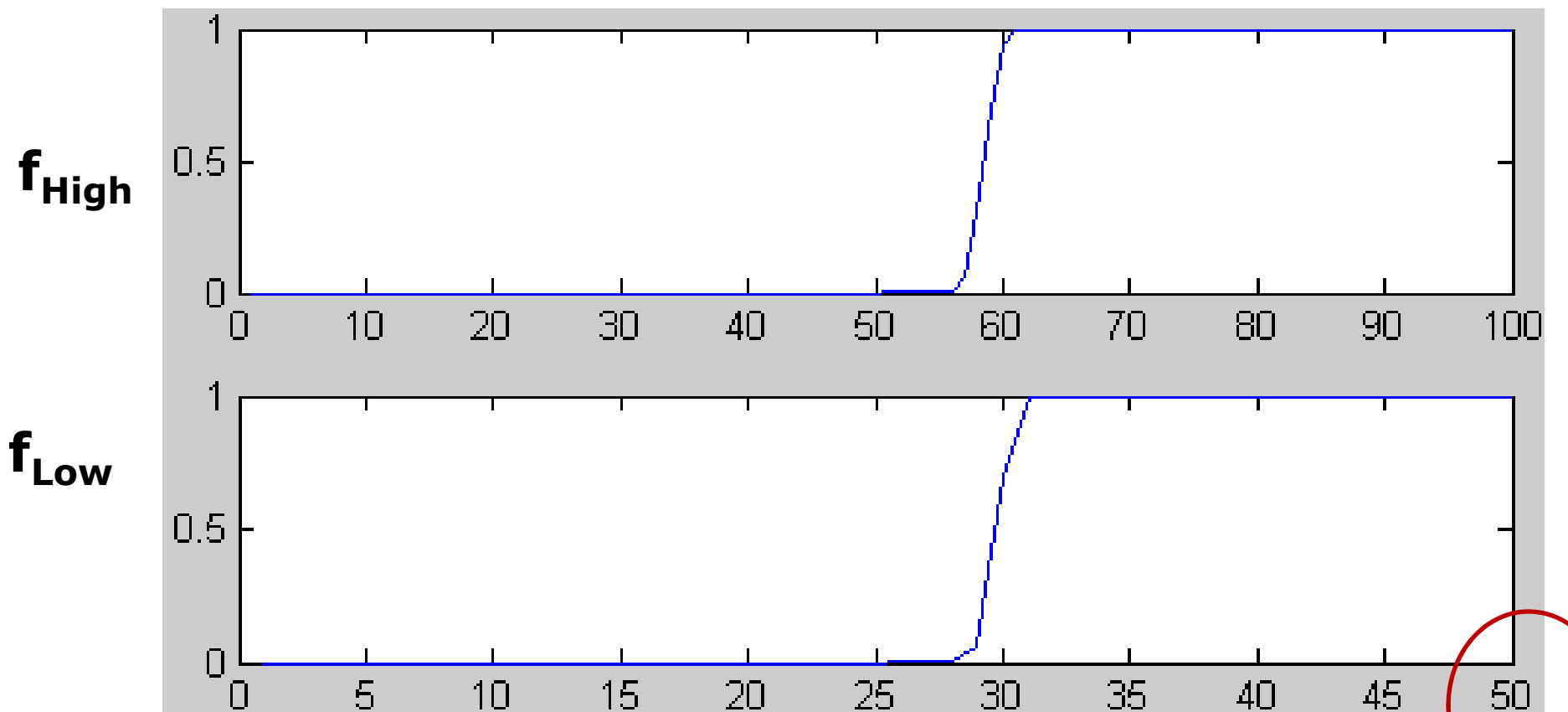
f_{High}



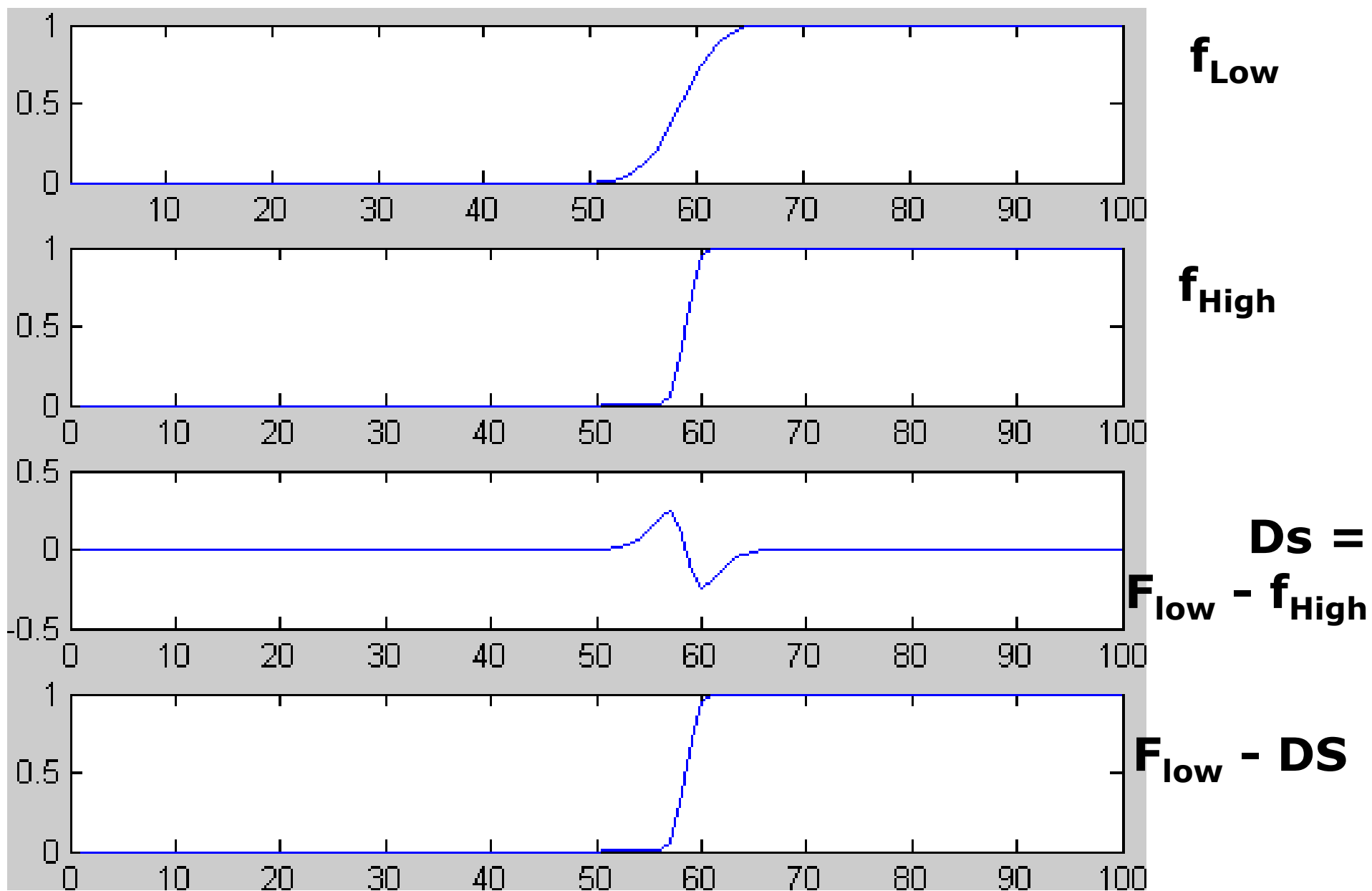
f_{Low}

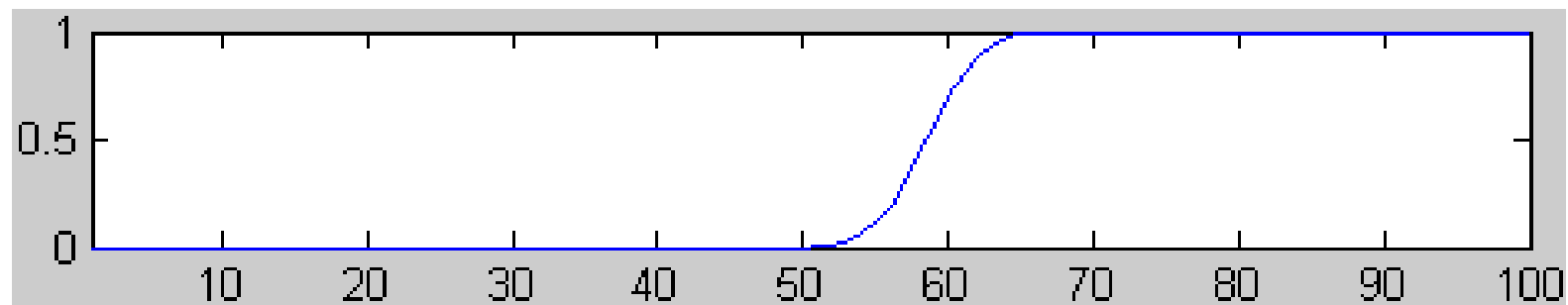


Upsampled (by 4) F_{Low}

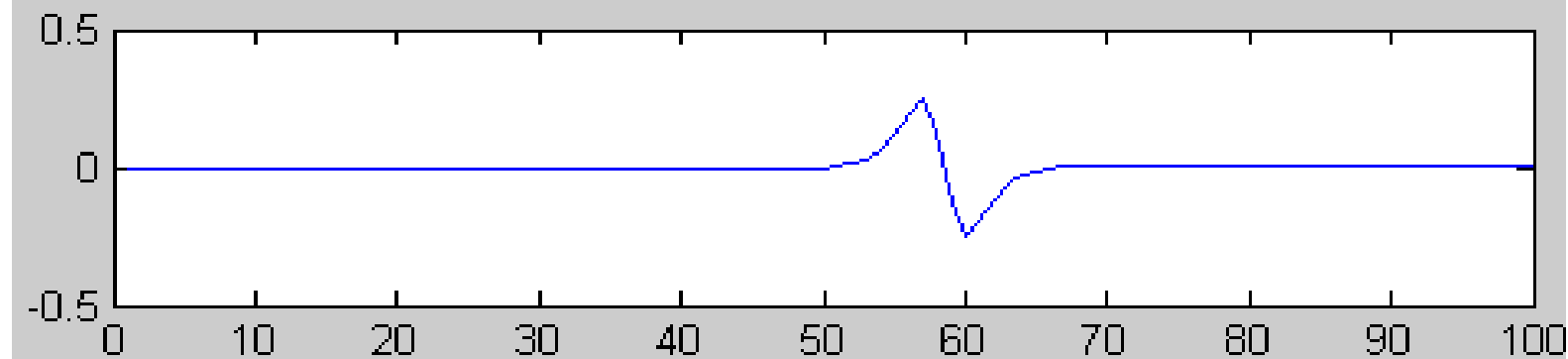


Upsampled (by 2) F_{low}

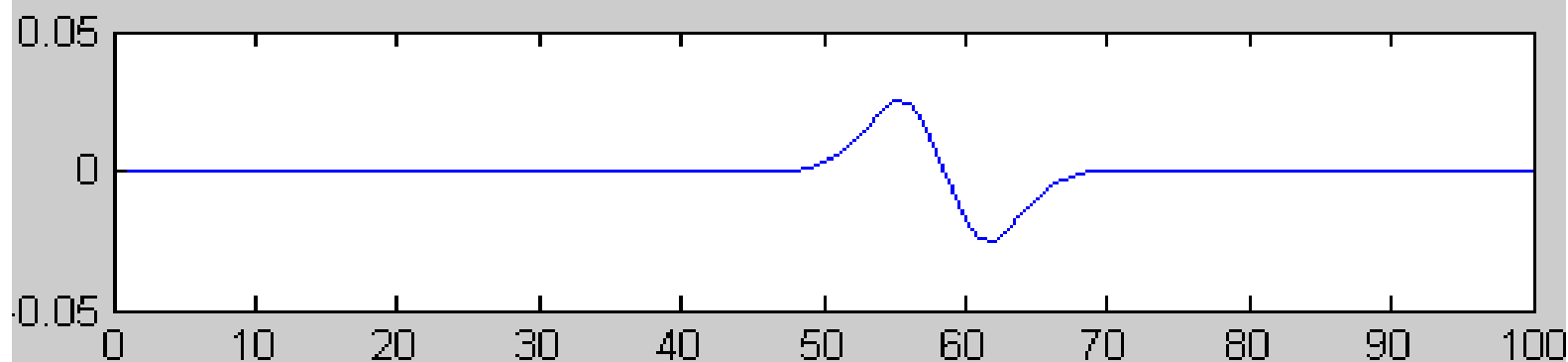




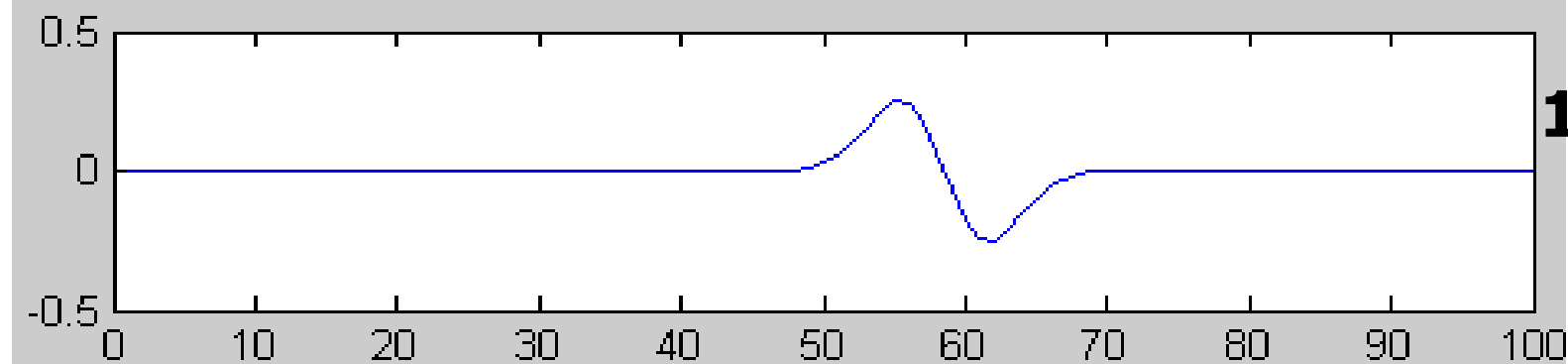
f_{Low}



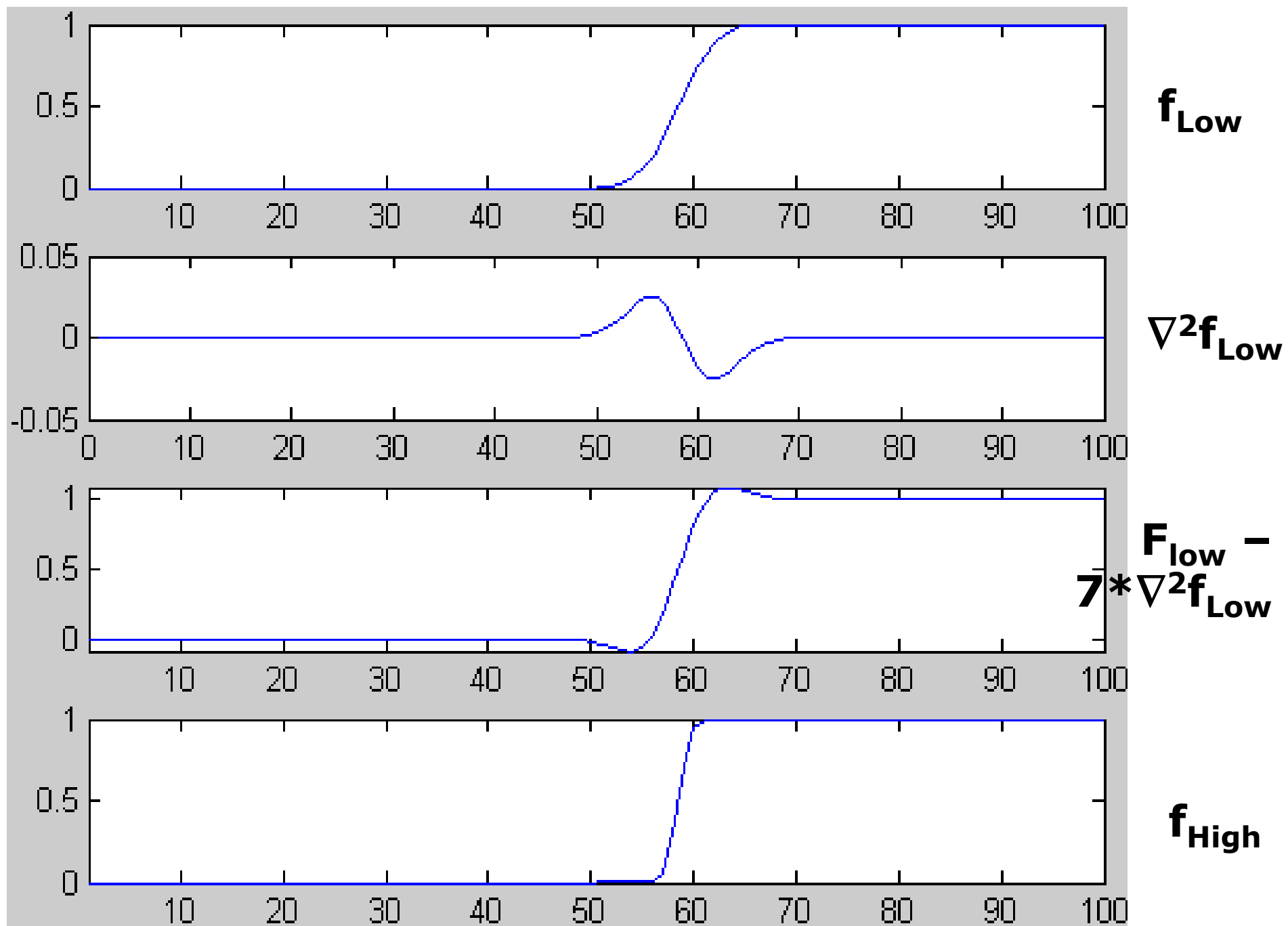
Ds



$\nabla^2 f_{\text{Low}}$



$10 * \nabla^2 f_{\text{Low}}$



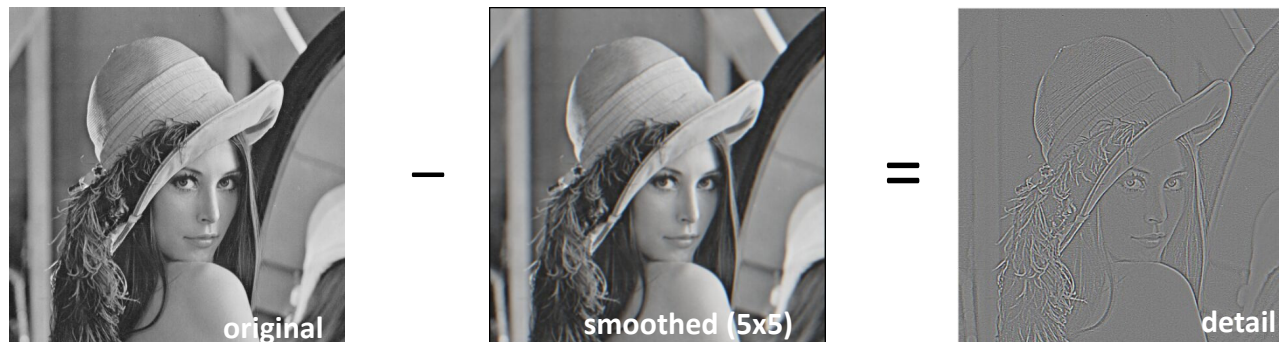
$$f_H = f_L - k * \nabla^2 f_L;$$

- **Need to learn an optimal K, for a class of edges;**
 - **Use del2(fL) as initial guess and formulate a signal re-shaping (anti-blur) or optimization,**
- or**
- work in transform domain ??**

Relates Enhancement of images →

Sharpening revisited

- What does blurring take away?

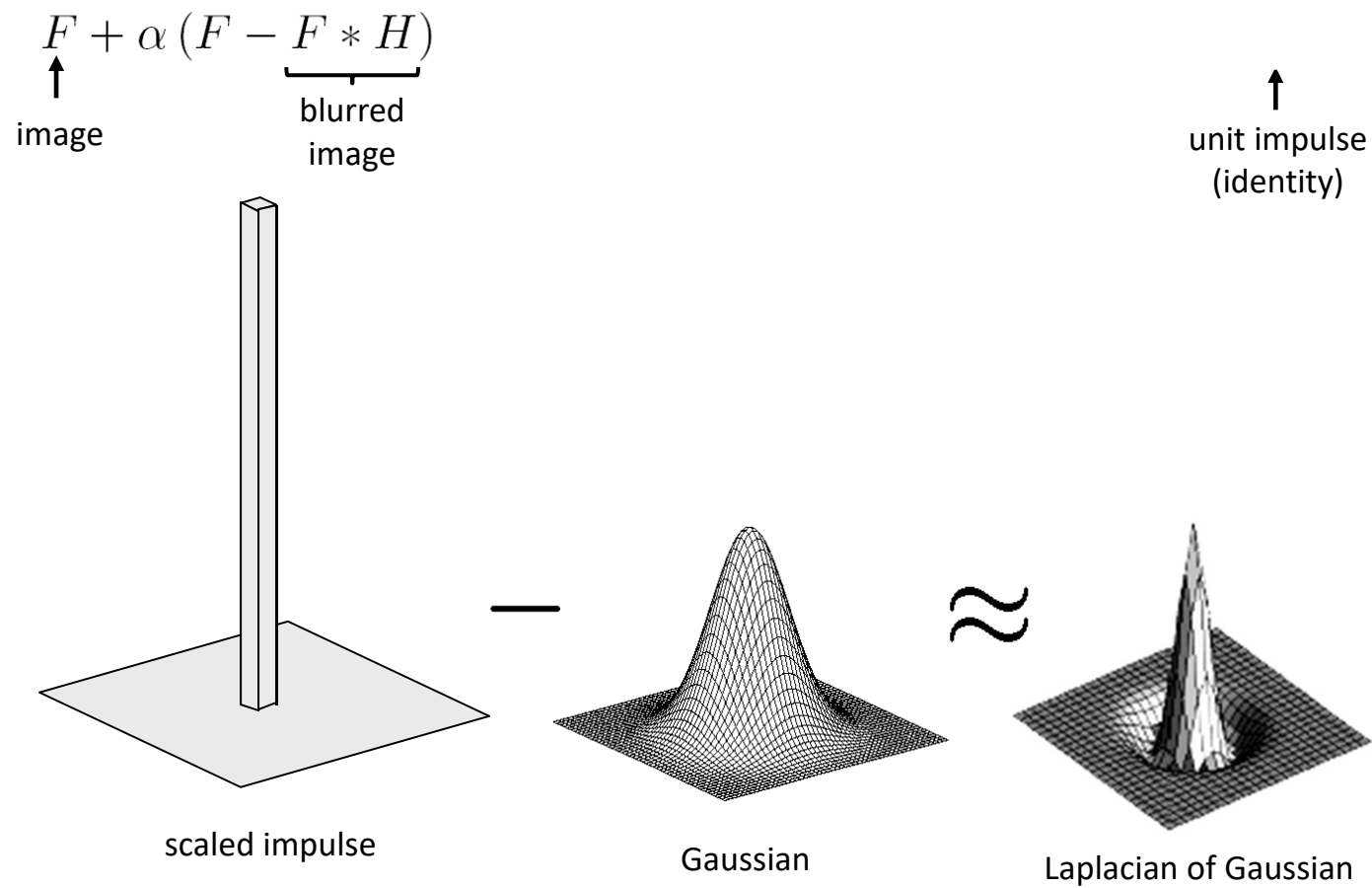


Let's add it back:



Source: S. Lazebnik

Sharpen filter



Sharpen filter



Unsharp Masking and Highboost Filtering Image enhancement can also be obtained using smoothing spatial filters, instead of the Laplacian. Assume that f is the input image (that may be blurry). We first compute a smoother image f_{smooth} by applying a smoothing linear filter to f as discussed before. Then the sharper output image g is defined by

$$g(x, y) = f(x, y) + k \cdot (f(x, y) - f_{smooth}(x, y)), \quad (6)$$

where $k \geq 0$ is a coefficient. When $k = 1$ the method is called unsharp masking. When $k > 1$, it is called highboost filtering, since the edges and details are even more highlighted.

Consider the case $k = 1$ in equation (6). When we compute $f - f_{smooth}$, this gives negative values on the left of the edge, and positive values on the right of the edge, and values close to 0 elsewhere. Thus when we compute $f + (f - f_{smooth})$ this provides the sharper corrected edge in g .

Consider first the denoising problem for simplicity. Let g be the given noisy image, f the image to be restored, linked through the relation $g(x, y) = f(x, y) + n(x, y)$.

Let's recall for a moment the enhancement method using the Laplacian: $g = f - \Delta f$, where f was the input blurry image and g was the sharper output image. However, to remove noise, we need the opposite process of smoothing. So let's invert the relation $g = f - \Delta f$ to obtain

$$f = g + \Delta f, \quad (11)$$

where g is the input noisy image and f is the denoised processed image. This is a linear partial differential equation, with unknown f . There are many ways to solve (11).

QUESTIONS AND DISCUSSIONS