

In [5]:

```
%matplotlib inline
import sys
sys.path.insert(0, '..')
from IPython.display import HTML, Image, SVG, YouTubeVideo
```

Image restoration

Image acquisition is rarely perfect, physics or external condition may deform the image being acquired, here are some example of typical problems:

Deformation model

The original image is $f(x, y)$ undergoes a deformation, given by H , and an additive noise $\eta(x, y)$ the acquired image is $g(x, y)$.

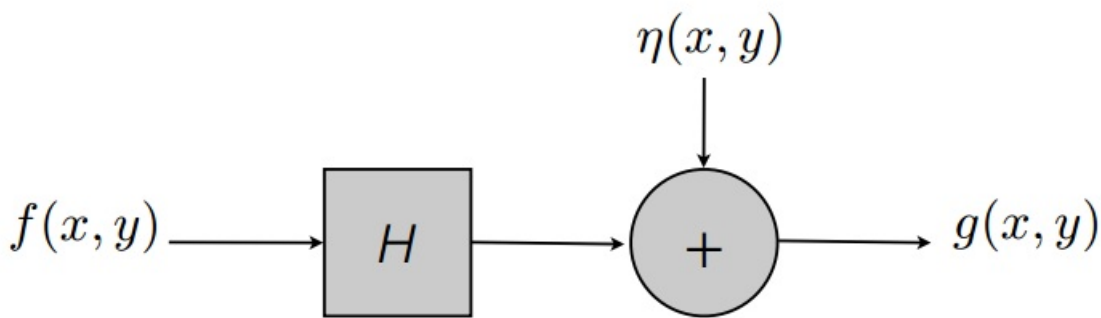
The restoration problem can be stated as follow:

how to recover a good approximation of $f(x, y)$ from $g(x, y)$?

In [6]:

```
Image( 'http://homepages.ulb.ac.be/~odebeir/data/restauration.png' )
```

Out[6]:



Some examples:

$$g(x, y) = H[f(x, y)] + \eta(x, y)$$

H properties:

- linear

$$H[k_1 f_1(x, y) + k_2 f_2(x, y)] = k_1 H[f_1(x, y)] + k_2 H[f_2(x, y)]$$

- additive

$$H[f_1(x, y) + f_2(x, y)] = H[f_1(x, y)] + H[f_2(x, y)]$$

- homogeneous

$$H[k_1 f_1(x, y)] = k_1 H[f_1(x, y)]$$

- spatially invariant

$$H[f(x - \alpha, y - \beta)] = g(x - \alpha, y - \beta)$$

Point Spread Function (PSF)

We can rewrite $f(x, y)$ as a sum of Dirac function.

$$f(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) \delta(x - \alpha, y - \beta) d\alpha d\beta$$

if there is no additive noise:

$$g(x, y) = H[f(x, y)] = H\left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) \delta(x - \alpha, y - \beta) d\alpha d\beta\right]$$

by linearity property:

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} H[f(\alpha, \beta) \delta(x - \alpha, y - \beta)] d\alpha d\beta$$

$f(\alpha, \beta)$ is independant of x and y :

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) H[\delta(x - \alpha, y - \beta)] d\alpha d\beta$$

the impulse response of H , also known as point spread function (PSF), is:

$$h(x, \alpha, y, \beta) = H[\delta(x - \alpha, y - \beta)]$$

after substitution we have:

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(x, \alpha, y, \beta) d\alpha d\beta$$

this expression means that, if the response H of an impulse is known, the response of any input $f(\alpha, \beta)$ is known.

if h is spatially invariant (see above):

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(x - \alpha, y - \beta) d\alpha d\beta$$

by adding the noise:

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(x - \alpha, y - \beta) d\alpha d\beta + \eta(x, y)$$

see also:

- { % cite gonzalez1977digital % } p254

Restoration

inverse filtering

If noise is negligible and PSF is known (in the Fourier domain):

$$\hat{F}(u, v) = G(u, v)H(u, v) \Rightarrow \hat{f}(x, y) = F^{-1}[\hat{F}(u, v)] = F^{-1}[G(u, v)H(u, v)]$$

else:

$$\hat{F}(u, v) = F(u, v) + N(u, v)H(u, v)$$

and noise is increasing when H is low, restoration is limited where H is big.

if define a restoration transform $M(u, v)$ as:

$$\hat{F}(u, v) = (G(u, v) + N(u, v))M(u, v)$$

one can use:

$$M(u, v) = \begin{cases} 1/H(u, v), & u^2 + v^2 \leq w_0^2 \\ 1, & u^2 + v^2 > w_0^2 \end{cases}$$

with w_0 being a distance to the origin in the Fourier space.

see also:

- { % cite rosenfeld1976digital % } p276

Wiener filtering

To avoid arbitrary setting of a parameter for the inverse transform such as w_0 , one can use the Wiener approach which consists in minimizing by least square some error function.

Without going into the details, the restoration transform is:

$$M(u, v) = \frac{H^*(u, v) |H(u, v)|^2}{|H(u, v)|^2 + S_{vv}(u, v) S_{gg}(u, v)}$$

where $H^*(u, v)$ is the complex conjugate of $H(u, v)$,

$S_{vv}(u, v)$ is the spectral density of the noise and

$S_{gg}(u, v)$ is the spectral density of the degraded image.

see also:

- {% cite sonka2014image %} p107

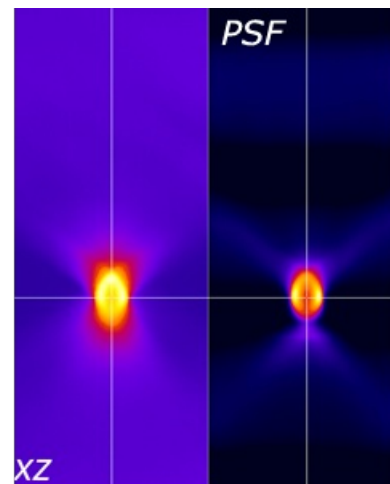
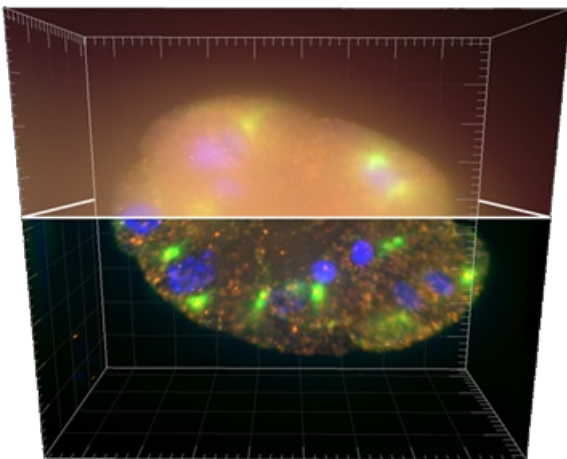
Blind deconvolution

If PSF is unknown, it has to be estimated. This is called blind deconvolution.

In [7]:

```
Image('http://bigwww.epfl.ch/algorithms/deconvolutionlab/meta/splash.png')
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Out[7]:



[image source \(http://bigwww.epfl.ch/algorithms/deconvolutionlab/\)](http://bigwww.epfl.ch/algorithms/deconvolutionlab/)

References

{% bibliography --cited %}