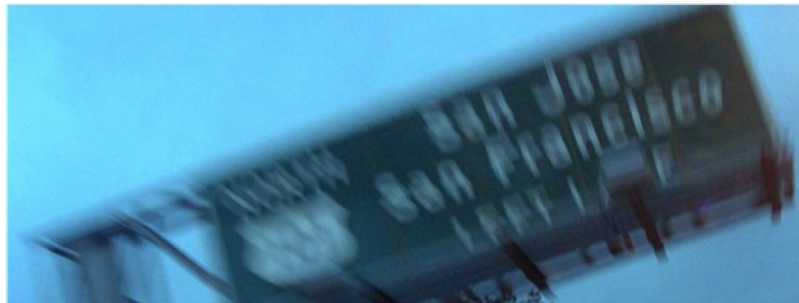


Lecture 12

Recall: Image deblurring



Atmospheric turbulence



Motion Blur



Speeding problem

Image deblurring in the frequency domain:

Mathematical formulation of image blurring

Let g be the observed (blurry) image.

Let f be the original (good) image.

$$\text{Model } g \text{ as: } g = H(f) + n$$

where H is the degradation function/operator and n is the additive noise.

Assumption on H :

1. H is position invariant:

$$\text{Let } g(x, y) = H(f)(x, y) \text{ and let } \tilde{f}(x, y) := f(x - \alpha, y - \beta).$$

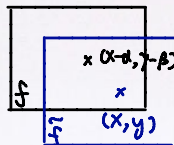
$$\text{Then: } H(\tilde{f})(x, y) = g(x - \alpha, y - \beta) = H(f)(x - \alpha, y - \beta)$$

2. Linear: $H(f_1 + f_2) = H(f_1) + H(f_2)$

$$H(\alpha f) = \alpha H(f) \text{ where } \alpha \text{ is a scalar multiplication.}$$

3. Linearity can be extended to integral:

$$H\left(\iint \alpha(u, v) f(x-u, y-v) du dv\right) = \iint \alpha(u, v) H(f)(x-u, y-v) du dv$$



With the above assumption, consider an impulse signal:

$$\delta(x, y) = \begin{cases} 1 & \text{if } (x, y) = (0, 0) \\ 0 & \text{if } (x, y) \neq (0, 0) \end{cases}$$

$$\text{Then: } f(x, y) = f * \delta(x, y) = \sum_{\alpha=-M/2}^{M/2-1} \sum_{\beta=-N/2}^{N/2-1} f(\alpha, \beta) \delta(x-\alpha, y-\beta)$$

$$\therefore g(x, y) = H(f)(x, y)$$

$$= \sum_{\alpha=-M/2}^{M/2-1} \sum_{\beta=-N/2}^{N/2-1} f(\alpha, \beta) H(\delta)(x-\alpha, y-\beta) \quad (\text{by linearity and position-invariant})$$

$$= \sum_{\alpha=-M/2}^{M/2-1} \sum_{\beta=-N/2}^{N/2-1} f(\alpha, \beta) h(x-\alpha, y-\beta) \quad \text{where } h(x, y) = H(\delta)(x, y)$$

$$= f * h(x, y)$$

\therefore With the above assumption,

Degradation/Blur = Convolution

Remark:

1. h is called the point spread function

2. $\therefore g(x,y) = h * f(x,y) + n(x,y)$

In the frequency domain,

$$G(u,v) = c H(u,v) F(u,v) + N(u,v)$$

\uparrow
constant

\therefore Deblurring can be done by:

$$\text{Compute: } F(u,v) \approx \frac{G(u,v)}{cH(u,v)}$$

\downarrow

— from observed image
— from known degradation

$$\text{Obtain: } f(x,y) = \text{DFT}^{-1}(F(u,v))$$

(Does NOT work very well due to noise!)

Examples of degradation function $H(u,v)$

1. Atmospheric turbulence blur:

$$H(u,v) = e^{-k(u^2+v^2)^{5/6}}$$

where k = degree of turbulence

$$k = 0.0025 \text{ (severe)}$$

$$k = 0.001 \text{ (mild)}$$

$$k = 0.00025 \text{ (low turbulence)}$$

2. Out of focus blur:

In the frequency domain, define $H(u,v)$ as the DFT of

$$h(x,y) = \begin{cases} 1 & \text{if } x^2 + y^2 \leq D_0^2 \\ 0 & \text{otherwise} \end{cases}$$

In some situations, a simple model:

$$H(u,v) = \begin{cases} 1 & \text{if } u^2 + v^2 \leq D_0^2 \\ 0 & \text{otherwise} \end{cases} \quad (\text{Usually not accurate})$$

3. Uniform Linear Motion Blur

Assume $f(x,y)$ undergoes planar motion during acquisition.

(original)

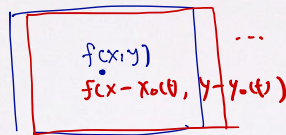
(displacements)

Let $(x_0(t), y_0(t))$ be the motion components in the x- and y-directions
↑
time

Let T be the total exposure time.

The observed image is given by:

$$g(x,y) = \int_0^T f(x-x_0(t), y-y_0(t)) dt$$



Now, let $G(u,v) = \text{DFT}(g)(u,v)$, then:

$$\begin{aligned} G(u,v) &= \frac{1}{N^2} \sum_x \sum_y g(x,y) e^{-j\frac{2\pi}{N}(ux+vy)} \\ &= \frac{1}{N^2} \sum_x \sum_y \int_0^T f(x-x_0(t), y-y_0(t)) dt e^{-j\frac{2\pi}{N}(ux+vy)} \\ &= \int_0^T \left(\sum_x \sum_y f(x-x_0(t), y-y_0(t)) e^{-j\frac{2\pi}{N}(ux+vy)} \right) dt \end{aligned}$$

Recall that $\text{DFT}(f(x-x_0, y-y_0)) = F(u, v) e^{-j\frac{2\pi}{N}(ux_0(t) + vy_0(t))}$

$$F = \text{DFT}(f)$$

We have: $G(u, v) = \int_0^T [F(u, v) e^{-j\frac{2\pi}{N}(ux_0(t) + vy_0(t))}] dt$

$$= F(u, v) \int_0^T e^{-j\frac{2\pi}{N}(ux_0(t) + vy_0(t))} dt$$

$$= F(u, v) H(u, v)$$

\therefore Degradation function in the frequency domain is given by:

$$H(u, v) = \int_0^T e^{-j\frac{2\pi}{N}(ux_0(t) + vy_0(t))} dt$$

(Speeding problem !!)

Example: Suppose the camera is moving left horizontally with a constant speed c .

That is, the image at time t is given by:

$$I^*(x, y) = I(x, y - ct)$$

Then: the degradation function is given by:

$$H(u, v) = \int_0^T e^{-j \frac{2\pi}{N} (v(ct))} dt$$

Remark: Once the degradation function is known, the original image can be restored by: $\text{IDFT}\left(\frac{G(u, v)}{H(u, v)}\right)$ (given that there's no noise)

What if there is noise??

Image deblurring in the frequency domain: (Assume H is known)

Method 1: Direct inverse filtering

$$\text{Let } T(u, v) = \frac{1}{H(u, v) + \varepsilon \operatorname{sgn}(H(u, v))} \quad (\operatorname{sgn}(z) = 1 \text{ if } \operatorname{Re}(z) \geq 0 \text{ and } \operatorname{sgn}(z) = -1 \text{ otherwise})$$

Avoid singularity

$$\text{Compute } \hat{F}(u, v) = G(u, v) T(u, v).$$

Find inverse DFT of $\hat{F}(u, v)$ to get an image $\hat{f}(x, y)$.

Good: Simple

Bad: Boost up noise

$$\hat{F}(u, v) = G(u, v) T(u, v) \approx F(u, v) + \frac{N(u, v)}{H(u, v) + \varepsilon \operatorname{sgn}(H(u, v))}$$

$H(u, v)F(u, v) + N(u, v)$

Note: $H(u, v)$ is big for (u, v) close to $(0, 0)$ (Keep low frequencies)
is small for (u, v) far away from $(0, 0)$

$\therefore \frac{N(u, v)}{H(u, v) + \varepsilon \operatorname{sgn}(H(u, v))}$ is big for (u, v) far away from $(0, 0)$

Large gain in high frequencies
↓

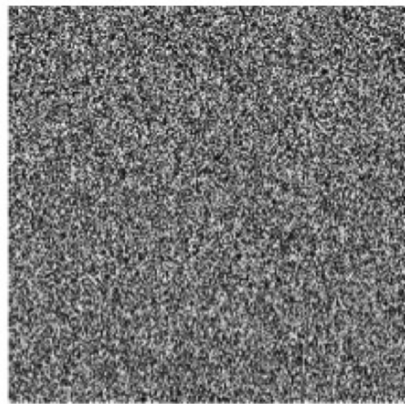
Boost up noises!!



Original



Blurred image



Direct inverse filtering