

OUTLINE

- 1) Reminder about SVD
- 2) Fourier (Freq. Domain) View
- 3) 2D Case

6.581/20.482
LECTURE 21: DECONVOLUTION II

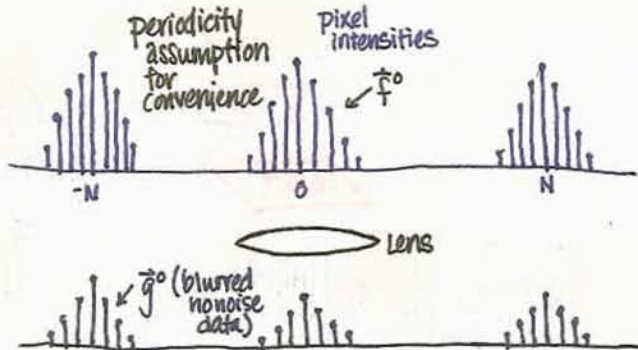
TUESDAY
2 MAY 2006

REMINDER object

$$\vec{g} = H \vec{f}^o + \vec{w}$$

↑ data
↑ effects of imaging system
↑ noise

IN 1D



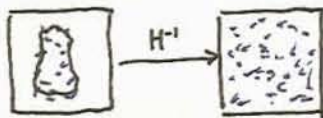
$$\begin{bmatrix} \hat{g}^o[1] \\ \vdots \\ \hat{g}^o[N] \end{bmatrix} = \begin{bmatrix} h[0] & h[N-1] & h[N-2] & \dots & h[2] & h[1] \\ h[1] & h[0] & h[N-1] & \dots & h[3] & h[2] \\ h[2] & h[1] & h[0] & \dots & h[4] & h[3] \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ h[N-1] & h[N-2] & h[N-3] & \dots & h[1] & h[0] \end{bmatrix} \begin{bmatrix} f^o[1] \\ \vdots \\ f^o[N] \end{bmatrix}$$

CIRCULANT MATRIX

1st DECONVOLUTION IDEA

$$\vec{f}^{BE} = H^{-1} \vec{g} = \vec{f}^o + H^{-1} \vec{w}$$

(bad estimate)



Using SVD

$$H = U \begin{bmatrix} \sigma_1 & \dots & \sigma_n \end{bmatrix} V^T$$

orthonormal
(define coordinate system)

$$\vec{g} = \sum \vec{u}_i \sigma_i (\vec{v}_i^T \vec{f}^o) + \vec{w}$$

$$\vec{f}^{BE} = \sum \frac{1}{\sigma_i} \vec{v}_i (\vec{u}_i^T \vec{g})$$

$H^{-1} \vec{g}$

$$\vec{f}^{BE} = \sum \vec{v}_i \left[(\vec{v}_i^T \vec{f}^o) + \frac{1}{\sigma_i} \vec{u}_i^T \vec{w} \right]$$

ESTIMATOR $\alpha_i = \text{either zero or one}$

$$\vec{f}^{SE} = \alpha_1 (\vec{v}_1^T \vec{f}^o + \frac{1}{\sigma_1} \vec{u}_1^T \vec{w}) \vec{v}_1 + \alpha_2 (\vec{v}_2^T \vec{f}^o + \frac{1}{\sigma_2} \vec{u}_2^T \vec{w}) \vec{v}_2 + \dots + \alpha_n (\vec{v}_n^T \vec{f}^o + \frac{1}{\sigma_n} \vec{u}_n^T \vec{w}) \vec{v}_n$$

(simple estimate) ↑ image ↑ noise

want to keep terms where image is much larger than the noise

$$\text{If } |\vec{v}_i^T \vec{f}^o| > \text{thresh} \left| \frac{1}{\sigma_i} \vec{u}_i^T \vec{w} \right| \Rightarrow \alpha_i = 1$$

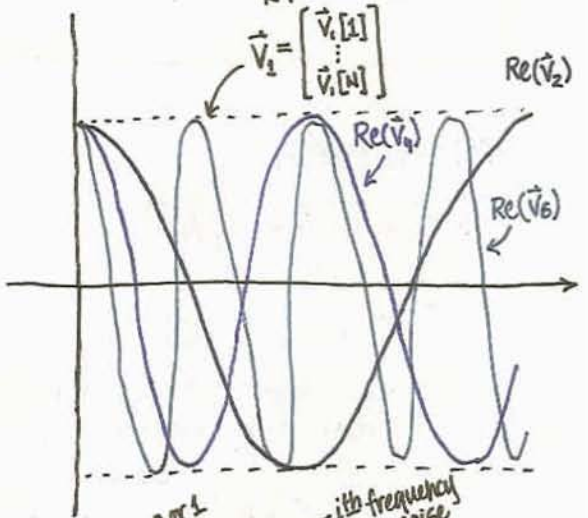
Else $\alpha_i = 0$

CONNECTION TO FREQUENCY RESPONSE

$$H_{\text{CIRCULANT}} = \begin{bmatrix} \text{DFT}^{-1} \end{bmatrix} \begin{bmatrix} \text{diagonal} \\ \text{Complex numbers} \end{bmatrix} \begin{bmatrix} \text{DFT} \\ \text{discrete fourier transform} \end{bmatrix}$$

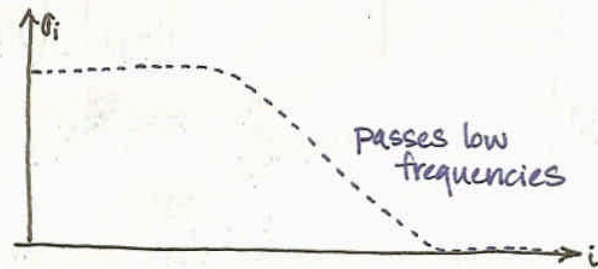
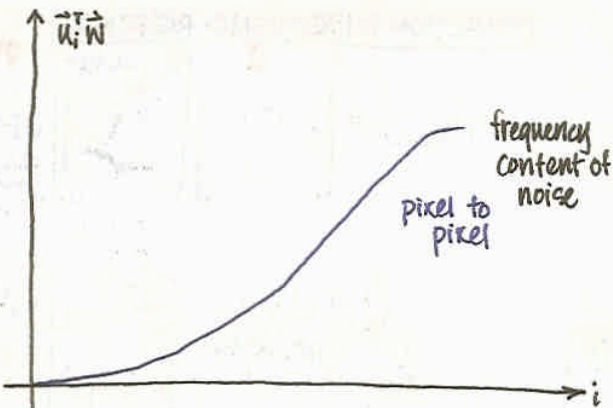
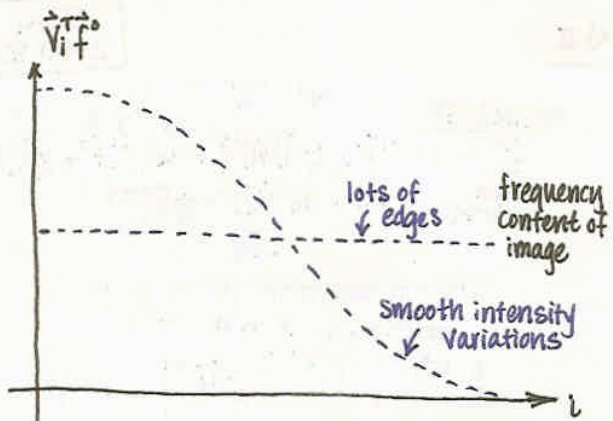
$$\frac{1}{N} \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ e^{j2\pi \frac{k-1}{N} (n-1)} \\ \vdots \\ \vdots \end{bmatrix} \begin{bmatrix} X[1] \\ \vdots \\ X[N] \end{bmatrix} = \begin{bmatrix} x[1] \\ \vdots \\ x[N] \end{bmatrix}$$

$$x[n] = \sum_{k=1}^N e^{j2\pi \frac{(k-1)(n-1)}{N}} X[k]$$

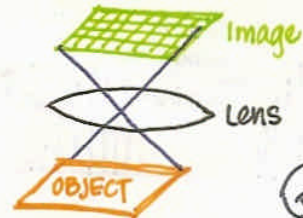
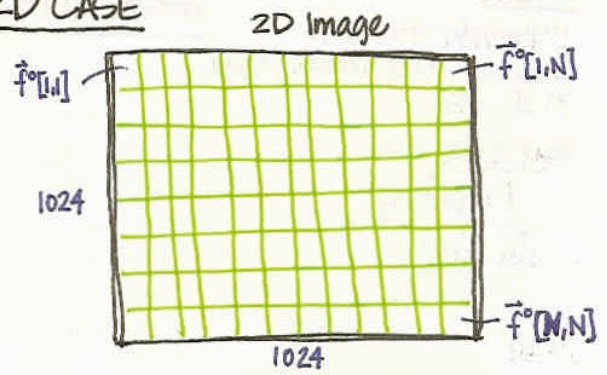


$$\vec{f}^{SE} = \sum \alpha_i (\vec{v}_i^T \vec{f}^o + \frac{1}{\sigma_i} \vec{u}_i^T \vec{w}) \vec{v}_i$$

↑ 0 or 1
↑ i-th frequency in noise
↑ i-th frequency in object



2D CASE

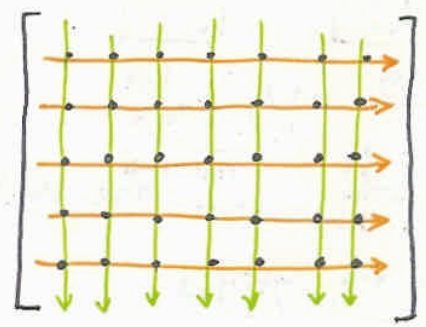


$$\begin{bmatrix} \tilde{g}^0[1,1] \\ \vdots \\ \tilde{g}^0[1,N] \\ \tilde{g}^0[2,1] \\ \tilde{g}^0[2,2] \\ \vdots \\ \tilde{g}^0[2,N] \\ \tilde{g}^0[3,1] \\ \vdots \\ \tilde{g}^0[N,N] \end{bmatrix} = \begin{bmatrix} \begin{matrix} H[0] & \dots & H[N-1] \\ \vdots & \ddots & \vdots \\ H[1] & \dots & H[N-1] \\ \vdots & \ddots & \vdots \\ H[N-1] & \dots & H[N-1] \end{matrix} \\ \vdots \\ \begin{matrix} H[N-1] \\ \vdots \\ H[1] \\ \vdots \\ H[0] \end{matrix} \end{bmatrix} \begin{bmatrix} F^0[1,1] \\ \vdots \\ F^0[1,N] \\ \vdots \\ F^0[N,N] \end{bmatrix}$$

1 million x 1 million

2D DFT (done fast with FFT)

$$X[n_1, n_2] = \frac{1}{N^2} \sum e^{-j2\pi \frac{(n_1-1)(k_1-1)}{N}} e^{-j2\pi \frac{(n_2-1)(k_2-1)}{N}} X[k_1, k_2]$$



Which terms to keep

$$\tilde{F}^E = \sum \alpha_i \tilde{V}_i (\tilde{V}_i^T \tilde{F}^0 + \frac{1}{\sigma_i} \tilde{u}_i \tilde{w})$$

Is α_i 1 or 0?

$$\tilde{V}_i^T \tilde{F}^0 \text{ versus } \frac{1}{\sigma_i} \tilde{u}_i \tilde{w}$$

Suppose we know the "energy" in the noise: $\tilde{w}^T \tilde{w} = E_{\text{noise}}$

& suppose we also know the "energy" in the image: $\tilde{F}^0^T \tilde{F}^0 = E_{\text{image}}$

\Rightarrow energy in the image is uniform in frequency

$$\tilde{F}^0 = \sum \tilde{V}_i (\tilde{V}_i^T \tilde{F}^0)$$

$$\tilde{F}^0^T \tilde{F}^0 = (\sum \tilde{V}_i (\tilde{V}_i^T \tilde{F}^0))^T (\sum \tilde{V}_i (\tilde{V}_i^T \tilde{F}^0))$$

$$|\tilde{V}_i^T \tilde{F}^0| = \frac{\sqrt{E_{\text{image}}}}{N}$$

Uniform Noise assumption: $|\tilde{u}_i \tilde{w}| = \frac{\sqrt{E_{\text{noise}}}}{N}$