

DIGITAL IMAGE PROCESSING

IMAGE RESTORATION

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IMAGE RESTORATION

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- Image degradation
- General image restoration models
- Inverse filtering
- Wiener filtering
- Constrained least squares filtering
- Geometric image transformation

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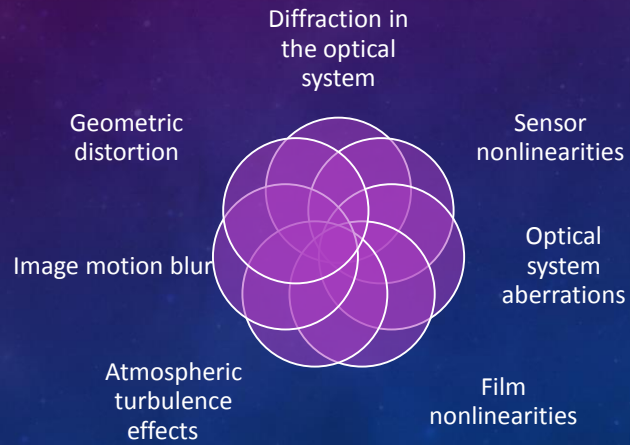
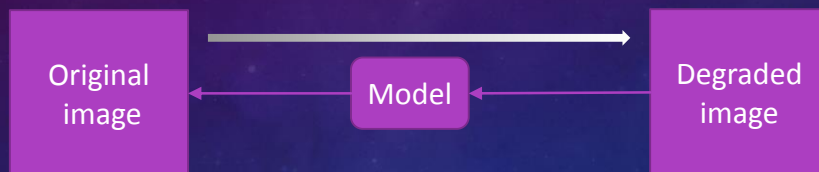


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- Goal



- Evaluation

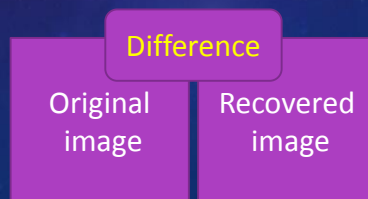


IMAGE RESTORATION

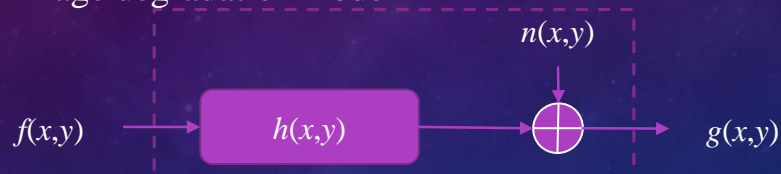
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- Traditional
 - Stable, linear position-invariant, priori knowledge of the signal and noise
- Modern
 - Non-stable(Kalman filter), nonlinear(ANN), no priori knowledge(blind image restoration)

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- General image degradation model



- Degraded output image in frequency domain

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

- Degraded output image in spatial domain

$$g(x,y) = h(x,y) \otimes f(x,y) + n(x,y)$$

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- Given $g(x, y)$, some knowledge about the degradation function H and some knowledge about the additive noise term $n(x, y)$, the objective of restoration is to obtain an estimate $\hat{f}(x, y)$ of the original image

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- Property of degradation function H
 - linear
 - Additivity
 - Homogeneity

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

- Position invariant

$$H[f(x-a, y-b)] = g(x-a, y-b)$$

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

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

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

additivity property

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

homogeneity property

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

- $h(x, \alpha, y, \beta)$ is called the *impulse response* of H , commonly referred to as the **point spread function**(PSF)
- The last line is called the superposition integral of the first kind
- H is completely characterized by its impulse response

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homogeneity property

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

position invariant

$$g(x, y) = f(x, y) \otimes h(x, y) + n(x, y)$$

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

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$$g = Hf + n$$

- $f(x, y)$ and $h(x, y)$ are of size $A \times B$, $C \times D$ respectively, by zero padding, they are extended to $M \times N$ ($M = A + C - 1$, $N = B + D - 1$) periodical functions:

$$f_e(x, y) = \begin{cases} f(x, y) & 0 \leq x \leq A-1, 0 \leq y \leq B-1 \\ 0 & \text{otherwise} \end{cases}$$

$$h_e(x, y) = \begin{cases} h(x, y) & 0 \leq x \leq C-1, 0 \leq y \leq D-1 \\ 0 & \text{otherwise} \end{cases}$$

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$$g_e(x, y) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} f_e(m, n) h_e(x-m, y-n) + n_e(x, y)$$

$$g = Hf + n$$

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- 2D Block Circulant Matrix

$$\mathbf{H} = \begin{bmatrix} \mathbf{H}_0 & \mathbf{H}_{M-1} & \mathbf{H}_{M-2} & \cdots & \mathbf{H}_1 \\ \mathbf{H}_1 & \mathbf{H}_0 & \mathbf{H}_{M-1} & \cdots & \mathbf{H}_2 \\ \mathbf{H}_2 & \mathbf{H}_1 & \mathbf{H}_0 & \cdots & \mathbf{H}_3 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \mathbf{H}_{M-1} & \mathbf{H}_{M-2} & \mathbf{H}_{M-3} & \cdots & \mathbf{H}_0 \end{bmatrix}$$

$$\mathbf{H}_j = \begin{bmatrix} h_e(j,0) & h_e(j,N-1) & h_e(j,N-2) & \cdots & h_e(j,1) \\ h_e(j,1) & h_e(j,0) & h_e(j,N-1) & \cdots & h_e(j,2) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ h_e(j,N-1) & h_e(j,N-2) & h_e(j,N-3) & \cdots & h_e(j,0) \end{bmatrix}$$

$$\mathbf{H} = \mathbf{W}\mathbf{D}\mathbf{W}^{-1}$$