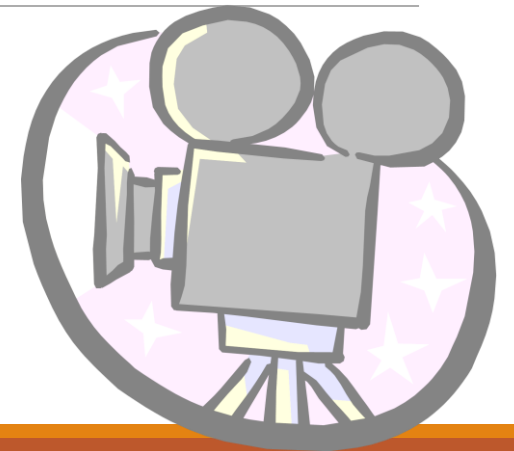


# Image Processing

CS-341

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# Outline

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- Image Enhancement in the Frequency Domain
  - Unitary Transformation
    - Fourier Transformation

# Transform=Change of Coordinates

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Intuitively speaking, transform plays the role of facilitating the source modeling

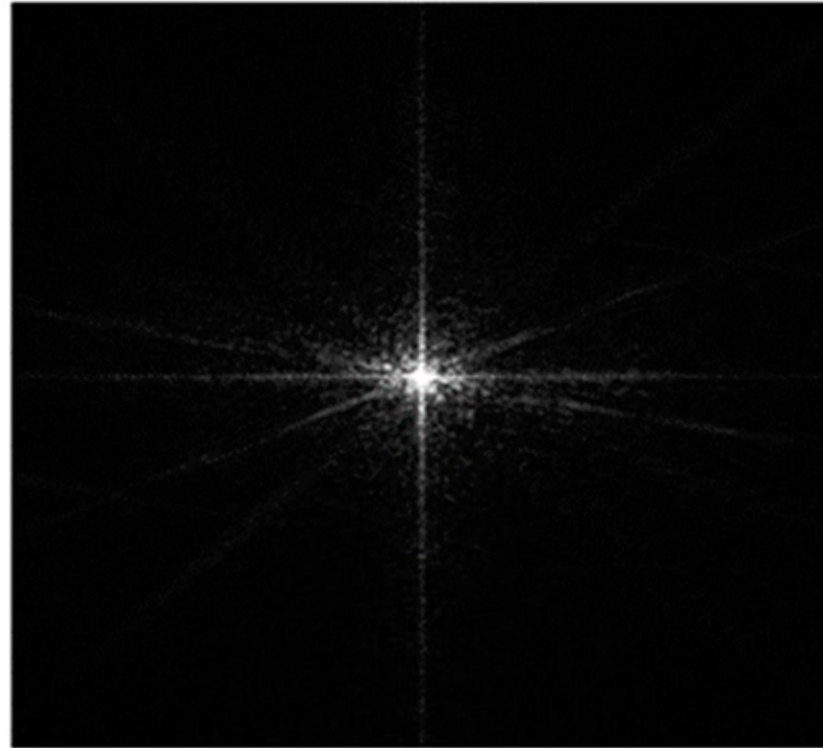
- Due to the decorrelating property of transform, it is easier to model transform coefficients (Y) instead of pixel values (X)

An appropriate choice of transform (transform matrix A) depends on the source statistics  $P(X)$

- We will only consider the **class of transforms** corresponding to **unitary matrices**

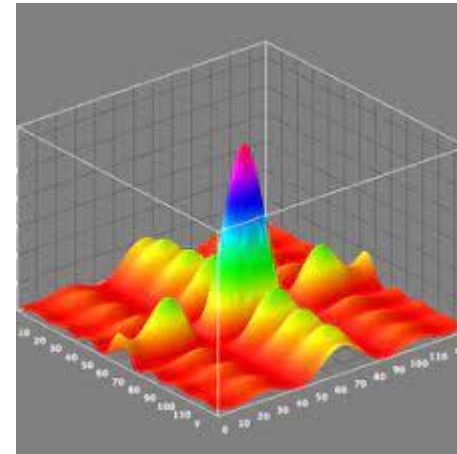
# image power circles

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Amplitude

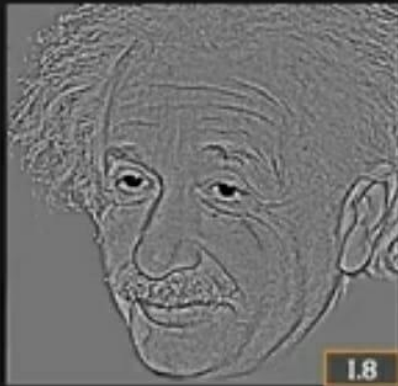


frequency

# Hybrid Images



Low Freq Only



High Freq Only



Hybrid (Sum) Image

# Unitary Matrix and 1D Unitary Transform

## Definition

A matrix  $A$  is called **unitary** if  $A^{-1} = A^*{}^T$

conjugate      transpose  
                 ↙      ↘

When the transform matrix  $A$  is unitary, the defined transform  $\vec{y} = A\vec{x}$  is called **unitary transform**

## Example

$$\mathbf{A} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}, \mathbf{A}^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} = \mathbf{A}^T$$

For a real matrix  $\mathbf{A}$ , it is unitary if  $\mathbf{A}^{-1} = \mathbf{A}^T$

# Properties of Unitary Transform

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Energy **compaction**: only few transform coefficients have large magnitude

- Such property is related to the decorrelating role of unitary transform

Energy **conservation**: unitary transform preserves the 2-norm of input vectors

- Such property essentially comes from the fact that rotating coordinates does not affect Euclidean distance



# Energy Compaction Example

Hadamard matrix

$$\mathbf{A} = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}, \vec{x} = \begin{bmatrix} 100 \\ 98 \\ 98 \\ 100 \end{bmatrix}$$

$$\vec{y} = \mathbf{A}\vec{x} = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 100 \\ 98 \\ 98 \\ 100 \end{bmatrix} = \begin{bmatrix} 198 \\ 0 \\ 0 \\ 2 \end{bmatrix}$$

significant  
insignificant

# Energy Conservation

$$\vec{y} = \mathbf{A}\vec{x} \quad \text{A is unitary}$$

$$\Rightarrow \|\vec{y}\|^2 = \|\vec{x}\|^2$$

$$\|\vec{y}\|^2 = \sum_{i=1}^N |y_i|^2$$

# Numerical Example

$$\mathbf{A} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}, \vec{x} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$



$$\vec{y} = \mathbf{A}\vec{x} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 7 \\ 1 \end{bmatrix}$$

Check:

$$\|\vec{x}\|^2 = 3^2 + 4^2 = 25, \|\vec{y}\|^2 = \frac{7^2 + 1^2}{2} = 25$$

# Summary of 1D Unitary Transform

Unitary matrix:  $\mathbf{A}^{-1} = \mathbf{A}^{*\top}$

Unitary transform:  $\vec{y} = \mathbf{A}\vec{x}$        $\mathbf{A}$  unitary

Properties of 1D unitary transform

- Energy compaction: most of transform coefficients  $y_i$  are small
- Energy conservation: quantization can be directly performed to transform coefficients

# Definition of 2D Transform

2D forward transform  $\mathbf{Y}_{N \times N} = \mathbf{A}_{N \times N} \mathbf{X}_{N \times N} \mathbf{A}_{N \times N}^T$

$$\begin{bmatrix} y_{11} & \cdots & \cdots & y_{1N} \\ \vdots & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ y_{N1} & \cdots & \cdots & y_{NN} \end{bmatrix} = \begin{bmatrix} a_{11} & \cdots & \cdots & a_{1N} \\ \vdots & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ a_{N1} & \cdots & \cdots & a_{NN} \end{bmatrix} \begin{bmatrix} x_{11} & \cdots & \cdots & x_{1N} \\ \vdots & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ x_{N1} & \cdots & \cdots & x_{NN} \end{bmatrix} \begin{bmatrix} a_{11} & \cdots & \cdots & a_{N1} \\ \vdots & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ a_{1N} & \cdots & \cdots & a_{NN} \end{bmatrix}$$

↑
↑

1D column transform
 1D row transform

# 2D Transform (Two Sequential 1D Transforms)

$$\mathbf{Y} = \mathbf{A}\mathbf{X}\mathbf{A}^T$$



column transform

$$\mathbf{Y}_1 = \mathbf{A}\mathbf{X} \quad (\text{left matrix multiplication first})$$

row transform

$$\mathbf{Y} = \mathbf{Y}_1\mathbf{A}^T = (\mathbf{A}\mathbf{Y}_1^T)^T$$



row transform

$$\mathbf{Y}_2 = \mathbf{X}\mathbf{A}^T = (\mathbf{A}\mathbf{X}^T)^T \quad (\text{right matrix multiplication first})$$

column transform

$$\mathbf{Y} = \mathbf{A}\mathbf{Y}_2$$

## Conclusion:

- 2D separable transform can be decomposed into two sequential
- The ordering of 1D transforms does not matter

# 2D Unitary Transform

Suppose  $\mathbf{A}$  is a unitary matrix,

forward transform

$$\mathbf{Y}_{N \times N} = \mathbf{A}_{N \times N} \mathbf{X}_{N \times N} \mathbf{A}_{N \times N}^T$$

inverse transform

$$\mathbf{X}_{N \times N} = \mathbf{A}_{N \times N}^{*T} \mathbf{Y}_{N \times N} \mathbf{A}_{N \times N}^*$$

Proof

Since  $\mathbf{A}$  is a unitary matrix, we have

$$\mathbf{A}^{-1} = \mathbf{A}^{*T}$$

$$\mathbf{A}^{*T} \mathbf{Y} \mathbf{A}^* = \mathbf{A}^{*T} (\mathbf{A} \mathbf{X} \mathbf{A}^T) \mathbf{A}^* = \mathbf{I} \cdot \mathbf{X} \cdot \mathbf{I} = \mathbf{X}$$

# Energy Compaction Property of 2D Unitary Transform

- Example

$$\mathbf{A} = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

$$\mathbf{X} = \begin{bmatrix} 100 & 100 & 98 & 99 \\ 100 & 100 & 94 & 94 \\ 98 & 97 & 96 & 100 \\ 100 & 99 & 97 & 94 \end{bmatrix} \xrightarrow{\mathbf{Y} = \mathbf{A}\mathbf{X}\mathbf{A}^T} \mathbf{Y} = \begin{bmatrix} 391.5 & 0 & 5.5 & 1 \\ 2.5 & -2 & -4.5 & 2 \\ 1 & -0.5 & 2 & -0.5 \\ 2 & 1.5 & 0 & -1.5 \end{bmatrix}$$

A coefficient is called **significant** if its magnitude is above a pre-selected threshold  $th$

insignificant coefficients ( $th=64$ )



# Example: Energy Compaction

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- Original Lena image



- 2D DCT



# Energy Conservation Property of 2D Unitary Transform

2-norm of a matrix  $\mathbf{X}$

$$\|\mathbf{X}\|^2 = \sum_{i=1}^N \sum_{j=1}^N |x_{ij}|^2$$

$$\mathbf{Y} = \mathbf{A}\mathbf{X}\mathbf{A}^T \quad \text{A unitary} \quad \longrightarrow \quad \|\mathbf{Y}\|^2 = \|\mathbf{X}\|^2$$

Example:

$$\mathbf{A} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$$

$$\mathbf{X} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \mathbf{Y} = \mathbf{A}\mathbf{X}\mathbf{A}^T \quad \longrightarrow \quad \mathbf{Y} = \begin{bmatrix} 5 & -1 \\ -2 & 0 \end{bmatrix}$$

$$\|\mathbf{X}\|^2 = 1^2 + 2^2 + 3^2 + 4^2 = 30 = 5^2 + 2^2 + 1^2 + 0^2 = \|\mathbf{Y}\|^2$$

# Suggested Readings

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- ❑ **Digital Image Processing by Rafael Gonzalez, Richard Woods, Pearson Education India, 2017.**
- ❑ **Fundamental of Digital image processing by A. K Jain, Pearson Education India, 2015.**

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Thank you

