

Lecture 4:

Haar transformation

Definition: (Haar functions) The Haar functions are defined recursively as follows

$$H_0(t) = \begin{cases} 1 & \text{if } 0 \leq t < 1 \\ 0 & \text{elsewhere.} \end{cases}$$

$$H_1(t) = \begin{cases} 1 & \text{if } 0 \leq t < \frac{1}{2} \\ -1 & \text{if } \frac{1}{2} \leq t < 1 \\ 0 & \text{elsewhere} \end{cases}$$

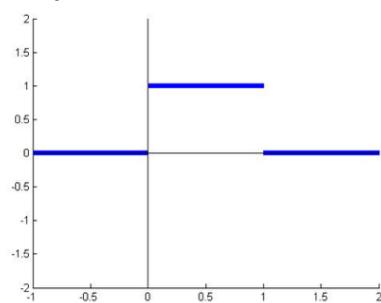
$$H_{2^p+n} = \begin{cases} \sqrt{2}^p & \text{if } \frac{n}{2^p} \leq t < \frac{n+0.5}{2^p} \\ -\sqrt{2}^p & \text{if } \frac{n+0.5}{2^p} \leq t < \frac{n+1}{2^p} \\ 0 & \text{elsewhere} \end{cases}$$

Where $p = 1, 2, \dots$; $n = 0, 1, 2, \dots, 2^p - 1$

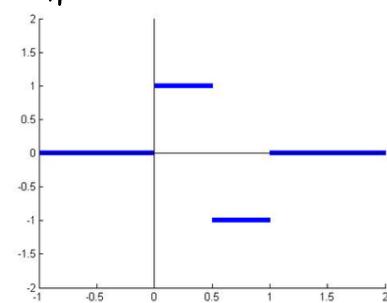
Remark: If p is larger, H_{2^p+n} is compactly supported in a smaller region.

Examples of Haar functions:

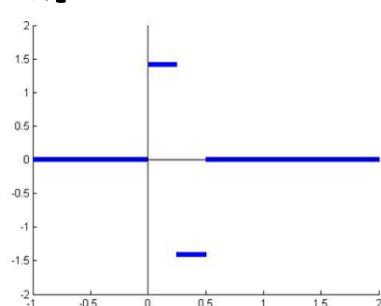
H_0



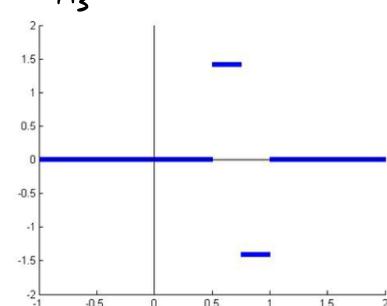
H_1



H_2



H_3



Definition (Discrete Haar Transform)

The Haar Transform of a $N \times N$ image is done by dividing $[0, 1]$ into partitions.

Let $H(k, i) = H_k\left(\frac{i}{N}\right)$ where $k, i = 0, 1, 2, \dots, N-1$.

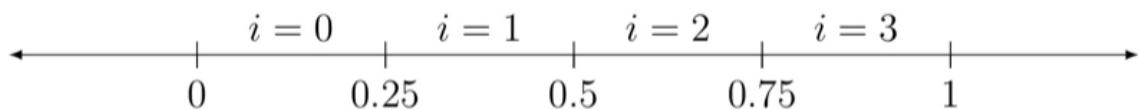
We obtain the Haar Transform matrix: $\tilde{H} = \frac{1}{\sqrt{N}} H$ where $H \equiv (H(k, i))_{0 \leq k, i \leq N-1}$

The Haar Transform of $f \in M_{n \times n}$ is defined as:

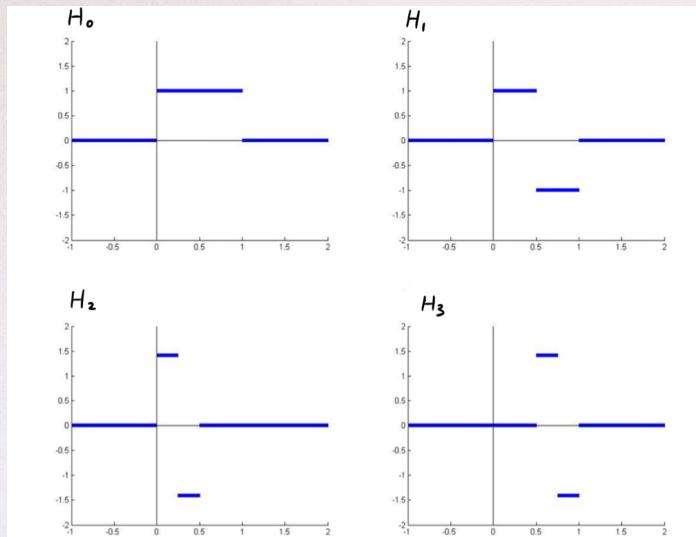
$$g = \tilde{H} f \tilde{H}^T.$$

Example Compute the Haar Transform matrix for a 4×4 image.

Solution: Divide $[0, 1]$ into 4 portions:



Need to check:



We get that:

$$H = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ \sqrt{2} & -\sqrt{2} & 0 & 0 \\ 0 & 0 & \sqrt{2} & -\sqrt{2} \end{pmatrix} \quad \text{and} \quad \tilde{H} = \frac{1}{\sqrt{4}} H = \frac{1}{2} H$$

Easy to check that $\tilde{H}^T \tilde{H} = I$.

Example 2 Compute the Haar Transform of

$$f = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$

Solution:

$$g = \tilde{H}f\tilde{H}^T = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 1 & -1 \end{pmatrix} \quad \left. \right\} \text{More zeros}$$

Example 3 Suppose g in Example 2 is changed to:

$$g = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

Reconstruct the original image.

Solution:

$$f = \tilde{H}^T f \tilde{H} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0.5 & 0.5 \\ 0 & 1 & 0 & 0 \end{pmatrix} \quad \text{Localized error}$$

Remark:

1. Haar Transform usually produces coefficient matrix with more zeros!

2. Localized error in coefficient matrix causes localized error in the reconstructed image

Elementary images under Haar transform:

Using Haar transform, f can be written as:

$$f = \tilde{H}^T g \quad \text{transformed image}$$

Let $\tilde{H} = \begin{pmatrix} \tilde{h}_1^T \\ \tilde{h}_2^T \\ \vdots \\ \tilde{h}_N^T \end{pmatrix}$. Then: $f = \sum_{i=1}^N \sum_{j=1}^N g_{ij} \tilde{h}_i \tilde{h}_j^T$

I_{ij}^T = elementary images under Haar Transform.

Definition: (Walsh function) The Walsh functions are defined recursively by:

$$W_{2j+q}(t) \equiv (-1)^{\lfloor \frac{j}{2} \rfloor + q} \{ W_j(2t) + (-1)^{j+q} W_j(2t-1) \}$$

where $\lfloor \frac{j}{2} \rfloor$ = biggest integer smaller than or equal to $\frac{j}{2}$.

$q = 0$ or 1 , $j = 0, 1, 2, \dots$ and

$$W_0(t) \equiv \begin{cases} 1 & 0 \leq t < 1 \\ 0 & \text{elsewhere} \end{cases}$$

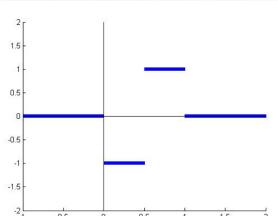
Example: Compute $W_1(t)$.

Put $j=0$, $q=1$. Then:

$$W_1(t) = (-1)^{\lfloor \frac{0}{2} \rfloor + 1} \{ W_0(2t) + (-1)^0 W_0(2t-1) \} = (-1) \{ W_0(2t) + (-1)^0 W_0(2t-1) \}$$

For $0 \leq t < \frac{1}{2}$, $W_0(2t) = 1$, $W_0(2t-1) = 0 \Rightarrow W_1(t) = -1$.

For $\frac{1}{2} \leq t < 1$, $W_0(2t) = 0$, $W_0(2t-1) = 1 \Rightarrow W_1(t) = 1$.



Definition: (Discrete Walsh transform)

The Walsh Transform of a $N \times N$ image is defined as follows.

Define $W(k, i) \equiv W_k\left(\frac{i}{N}\right)$ where $k, i = 0, 1, 2, \dots, N-1$.

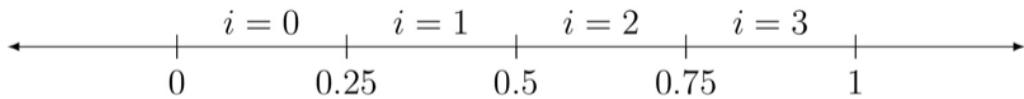
The Walsh transform matrix is: $\tilde{W} \equiv \frac{1}{\sqrt{N}} W$ where $W \equiv (W(k, i))_{0 \leq k, i \leq N-1}$

The Walsh transform of $f \in M_{n \times n}$ is defined as:

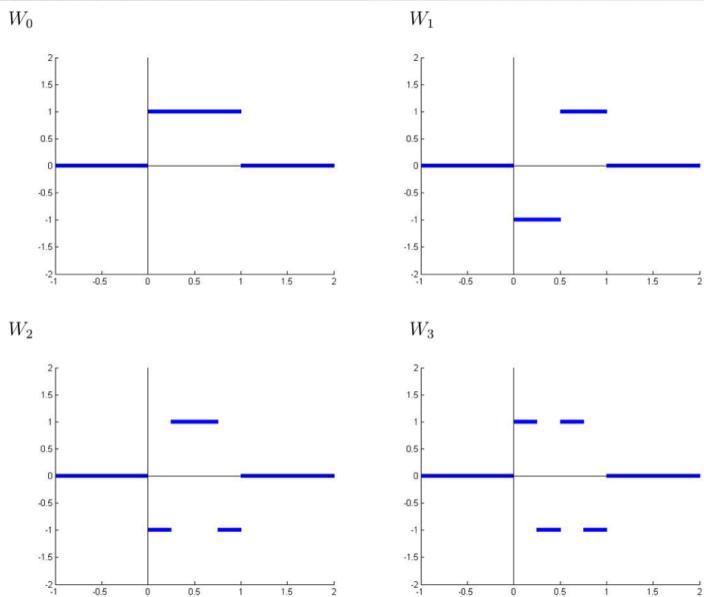
$$g = \tilde{W} f \tilde{W}^T$$

Example Compute the Walsh Transform matrix for a 4×4 image.

Solution: Again, divide $[0, 1]$ into 4 portions:



We can check that:



So,

$$W = \begin{pmatrix} 1 & 1 & 1 & 1 \\ -1 & -1 & 1 & 1 \\ -1 & 1 & 1 & -1 \\ 1 & -1 & 1 & -1 \end{pmatrix} \quad \text{and} \quad \tilde{W} = \frac{1}{\sqrt{4}}W = \frac{1}{2}W$$

$$(\tilde{W}^T \tilde{W} = I)$$

Example 2.7: Compute the Walsh Transform of

$$f = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$

Solution:

$$g = \tilde{W} f \tilde{W}^T = \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad \left. \right\} \text{More zeros in the coefficient matrix!}$$

Remark: 1. Walsh transform is to transform an image to a "transformed image" with much more zeros.

Elementary images under Walsh transform:

Under Walsh Transform, $f = \tilde{W}^T g \tilde{W}$.

Then: $f = \sum_{i=1}^N \sum_{j=1}^N g_{ij} \tilde{W}_i \tilde{W}_j^T$ where $\tilde{W} = \begin{pmatrix} -\tilde{w}_1^T & - \\ -\tilde{w}_2^T & - \\ \vdots & \\ -\tilde{w}_N^T & - \end{pmatrix}$

\tilde{W}_{ij} = elementary images under Walsh transform.

Walsh functions and sine function

Definition: (Rademacher function)

A Rademacher function of order n ($n \neq 0$) is defined as :

$$R_n(t) \equiv \text{sign}[\sin(2^n \pi t)] \text{ for } 0 \leq t \leq 1.$$

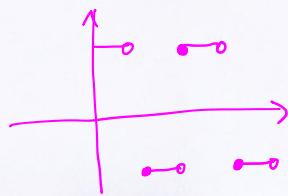
Where $\text{sign}(x) = 1$ if $x > 0$, $\text{sign}(x) = -1$ if $x < 0$ and $\text{sign}(x) = 0$ if $x = 0$.

For $n=0$, $R_0(t) \equiv 1$ for $0 \leq t \leq 1$.

Let $N = b_{m+1} 2^m + b_m 2^{m-1} + \dots + b_1 2^0$. Then, the R-Walsh function \tilde{W}_N is given by:

$$\tilde{W}_N = \prod_{i=1, b_i \neq 0}^{m+1} R_i(t)$$

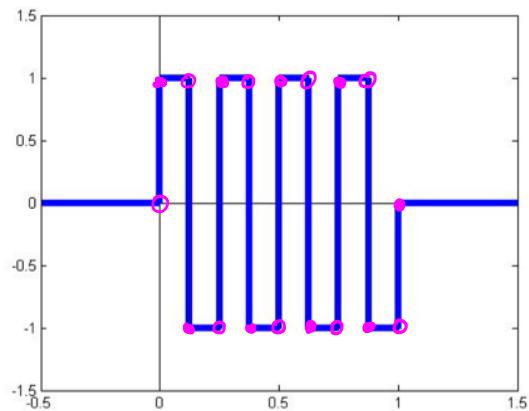
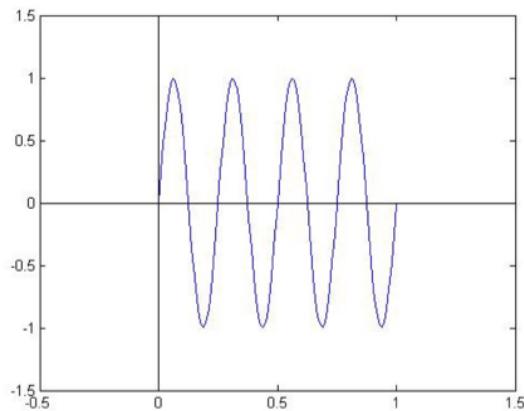
(where the values at the jumps are defined such that the function is continuous from the right)



Example : Compute R-Walsh function \tilde{W}_4 using Rademacher function.

Consider $\sin(8\pi t)$:

Therefore, $R_3(t) =$



As $4 = \underbrace{1}_{b_3} \cdot 2^2 + \underbrace{0}_{b_2} \cdot 2^1 + \underbrace{0}_{b_1} \cdot 2^0$, we have

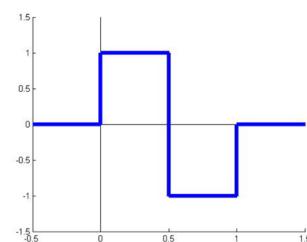
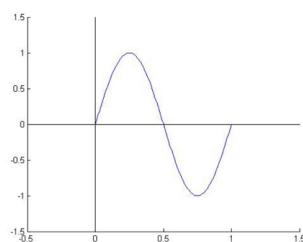
$$\tilde{W}_4 = \prod_{i=1, b_i \neq 0}^3 R_i(t) = R_3(t)$$

$$(W_{2j+q}(t) \equiv (-1)^{\lfloor j/2 \rfloor + q} \{W_j(2t) + (-1)^{j+q} W_j(2t-1)\})$$

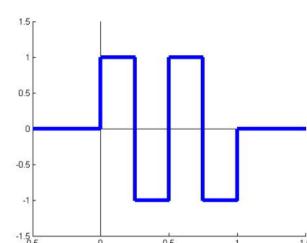
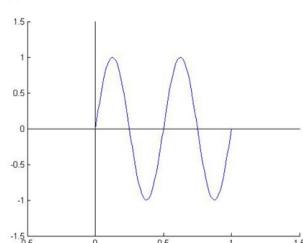
For $W_3(t)$: As $3 = \underbrace{1}_{b_2} \cdot 2^1 + \underbrace{1}_{b_1} \cdot 2^0$, we have

$$\tilde{W}_3(t) = \prod_{i=1, b_i \neq 0}^2 R_i(t) = R_1(t)R_2(t)$$

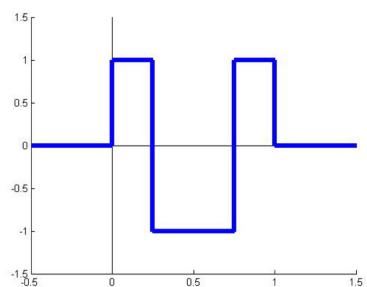
$R_1(t)$:



$R_2(t)$:



Therefore, $\tilde{W}_3(t)$:



Relationship between Walsh functions and R-Walsh functions

$$W_0(t) = \tilde{W}_0(t), W_1(t) = -\tilde{W}_1(t), W_2(t) = -\tilde{W}_3(t), W_3(t) = \tilde{W}_2(t), \\ W_4(t) = \tilde{W}_6(t), W_5(t) = -\tilde{W}_7(t), W_6(t) = -\tilde{W}_5(t), W_7(t) = \tilde{W}_4(t)$$

But how to get these formula?



How to determine Walsh from R -Walsh?

Write $i = b_{m+1} 2^m + b_m 2^{m-1} + \dots + b_1 2^0$.

Determine j whose binary representation is given by:
 $c_{m+1} c_m \dots c_1$ where

$$c_{m+1} = b_{m+1} \pmod{2}, \quad c_k = (b_{k+1} + b_k) \pmod{2}$$

Then: $W_i(t) = \pm \tilde{W}_j(t)$

The \pm sign is determined from $W_i(0)$!

Example 2.9 Consider $W_7(t)$.

Check that $W_7(t) > 0$.

Now, $7 = 2^2 + 2^1 + 2^0$ so binary representation of 7 is 111.

Therefore, $j = 100$ (binary) = 4

Thus, $W_7(t) = \tilde{W}_4(t)$.

Discrete Fourier Transform:

Definition: The 1D discrete Fourier Transform (DFT) of a function $f(k)$, defined at discrete points $k=0, 1, 2, \dots, N-1$ is defined as:

$$\hat{f}(m) = \frac{1}{N} \sum_{k=0}^{N-1} f(k) e^{-j2\pi m k} \quad (\text{where } j = \sqrt{-1}, e^{j\theta} = \cos \theta + j \sin \theta)$$

The 2D DFT of a $M \times N$ image $g = (g(k, l))_{k,l}$, where $0 \leq k \leq M-1$, $0 \leq l \leq N-1$ is defined as:

$$\hat{g}(m, n) = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi \left(\frac{km}{M} + \frac{ln}{N} \right)}$$

Remark: The inverse of DFT is given by:

$$g(p, q) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} \hat{g}(m, n) e^{j2\pi \left(\frac{pm}{M} + \frac{qn}{N} \right)}$$

↑ ↑ ↑
 (no $\frac{1}{MN}!$) DFT of g (no -ve sign)

Proof of Inverse DFT:

$$\begin{aligned}
 & \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} e^{j2\pi(\frac{pm}{M} + \frac{qn}{N})} \hat{g}(m, n) = \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} e^{j2\pi(\frac{pm}{M} + \frac{qn}{N})} \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi(\frac{km}{M} + \frac{ln}{N})} \\
 & = \frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} g(k, l) e^{j2\pi((\frac{p-k}{M})m + (\frac{q-l}{N})n)} \\
 & = \underbrace{\frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l)}_{(*)} \sum_{m=0}^{M-1} e^{j2\pi(\frac{(p-k)m}{M})} \sum_{n=0}^{N-1} e^{j2\pi(\frac{(q-l)n}{N})}
 \end{aligned}$$

Note that: $\sum_{m=0}^{M-1} e^{j2\pi(\frac{mt}{M})} = \frac{[e^{j2\pi(\frac{t}{M})}]^M - 1}{e^{j2\pi(\frac{t}{M})} - 1} = M \delta(t) := \begin{cases} M & t=0 \\ 0 & t \neq 0 \end{cases}$

$\therefore (*)$ becomes: $\frac{1}{MN} \sum_{k=0}^{M-1} \sum_{l=0}^{N-1} g(k, l) M \delta(p-k) N \delta(q-l) = g(p, q).$

Image decomposition under DFT:

Consider a $N \times N$ image g , the DFT of g :

$$\hat{g}(m, n) = \frac{1}{N^2} \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} g(k, l) e^{-j2\pi \left(\frac{km+ln}{N} \right)}$$

Define $U_{kl} = \frac{1}{N} e^{-j \frac{2\pi k l}{N}}$ where $0 \leq k, l \leq N-1$ and $U = (U_{kl})_{0 \leq k, l \leq N-1} \in M_{N \times N}$

U is clearly symmetric and also:

$$\hat{g} = U g U \quad (\text{DFT in matrix form})$$

$$\begin{aligned} \text{Note that: } \frac{1}{N^2} \sum_{\alpha=0}^{N-1} e^{-j \frac{2\pi x_1 \alpha}{N}} e^{+j \left(\frac{2\pi x_2 \alpha}{N} \right)} &= \frac{1}{N^2} \sum_{\alpha=0}^{N-1} e^{-j \frac{2\pi (x_2 - x_1) \alpha}{N}} = \frac{1}{N^2} N \delta(x_2 - x_1) \\ &= \frac{1}{N} \delta(x_2 - x_1) \end{aligned}$$

Let $U = \begin{pmatrix} \vec{u}_1 & \vec{u}_2 & \dots & \vec{u}_N \end{pmatrix}$. Then: $\langle \vec{u}_i, \vec{u}_j \rangle = \vec{u}_i^\top \vec{u}_j = \begin{cases} \frac{1}{N} & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$

$\therefore \{\vec{u}_i\}_{i=1}^N$ is orthogonal but NOT orthonormal!

$$\therefore UU^* = \frac{1}{N} I = U^*U$$

$$\therefore g = (NU)^* \hat{g} (NU)^*$$

$$\therefore g = \sum_{k=0}^{N-1} \sum_{l=0}^{N-1} \hat{g}_{kl} \vec{\omega}_k \vec{\omega}_l^T \quad \text{Elementary image of DFT}$$

where $\vec{\omega}_k = k^{\text{th}} \text{ col of } (NU)^*$

$$\hat{g} = U g U$$

$$\Rightarrow U^* \hat{g} U^* = (U^* U) g (U U^*)$$

$$\therefore (NU)^* \hat{g} (NU)^* = \left(\frac{1}{N}\right) g \left(\frac{1}{N}\right)$$

Example Find the DFT of the following 4×4 image

$$g = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

Solution

The matrix U is given by:

$$U = \frac{1}{4} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{pmatrix} \quad U = \left(u_{k,l} \right)_{k,l}$$
$$\therefore \text{DFT of } g = \hat{g} = UgU = \left(e^{-j2\pi \left(\frac{k+l}{4} \right)} \right)$$