

P.1.

(a) (20%) Eq. (3.5-20 a,b,c)

$$\begin{bmatrix} I \\ v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 1/3 & 1/3 & 1/3 \\ -1/\sqrt{6} & -1/\sqrt{6} & 2/\sqrt{6} \\ 1/\sqrt{6} & -1/\sqrt{6} & 0 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

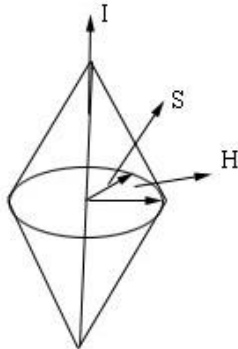
$$H = \tan^{-1}\{v_2 / v_1\}$$

$$S = (v_1^2 + v_2^2)^{1/2}$$

Using the above equations, we get

	Red	Yellow	Blue	Ref. White
I	1/3	2/3	1/3	1
H	$-\mathbf{p}/4$	0	0	0
S	$1/\sqrt{3}$	$2/\sqrt{6}$	$2/\sqrt{6}$	0

(b) (14%)



One possible distance measure between colors in HSI is

$$D = \mathbf{a}|I_1 - I_2| + \mathbf{b}|S_1 - S_2| + \mathbf{g}|H_1 - H_2|$$

$\mathbf{a}, \mathbf{b}, \mathbf{g}$ are weights controlling the relative importance among different channels.

Another possible measure is L_2 distance

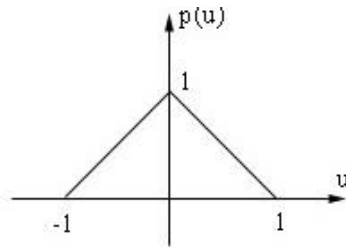
$$D = \left(|I_1 - I_2|^2 + |S_1 - S_2|^2 + |H_1 - H_2|^2 \right)^{1/2}$$

but such metrics may not correspond well to perceived distance.

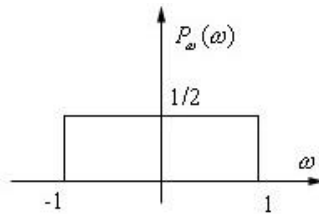
P.2.

(a) (17%)

$$p_u(u) = \begin{cases} 1 - |u| & -1 \leq u \leq 1 \\ 0 & \text{Otherwise} \end{cases}$$



We want to design $w = f(u)$ so that pdf of w is as follows:



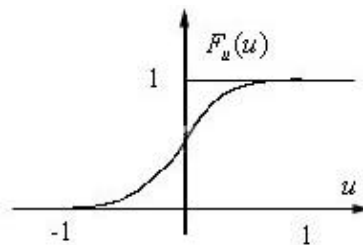
As described in the class notes,

$F_w(w) = F_u(u)$, where F is the cumulative probability distribution function

$$\frac{w - w_{\min}}{w_{\max} - w_{\min}} = F_u(u), \quad w_{\min} \leq w \leq w_{\max} \quad (*)$$

Now $w_{\min} = -1$, $w_{\max} = 1$

$$F_u(u) = \int_{-\infty}^u p_u(x) dx = \begin{cases} (u+1)^2 / 2 & -1 \leq u \leq 0 \\ -u^2 / 2 + u + 1/2 & 0 \leq u \leq 1 \end{cases}$$

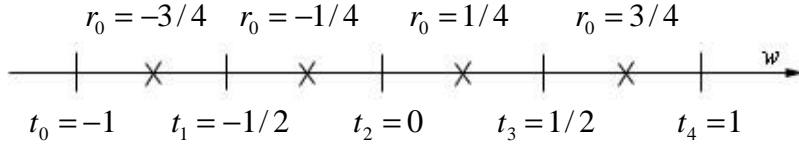


Use equation (*) above,

$$w = 2F_u(u) - 1 = \begin{cases} u^2 + 2u, & -1 \leq u \leq 0 \\ -u^2 + 2u, & 0 \leq u \leq 1 \end{cases} \quad (**)$$

(b) (17%)

Since the pdf of w is uniform, the optimal quantization for w will be the uniform quantizer.

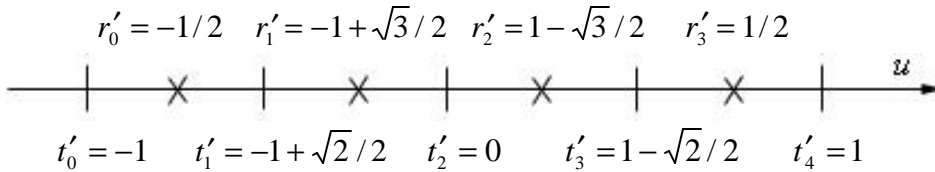


Map t_i & r_i for w to the decision and reconstruction values for u .
Use equation (***) in part (a) to derive the reverse mapping function

$$u = \begin{cases} -1 + \sqrt{1+w} & -1 \leq w \leq 0 \\ 1 - \sqrt{1+w} & 0 \leq w \leq 1 \end{cases}$$

\therefore

$$\begin{aligned} t_0 = -1 &\Rightarrow t'_0 = -1 \\ t_1 = -1/2 &\Rightarrow t'_1 = -1 + \sqrt{1/2} \\ t_2 = 0 &\Rightarrow t'_2 = 0 \\ t_3 = 1/2 &\Rightarrow t'_3 = 1 - \sqrt{1/2} \\ t_4 = 1 &\Rightarrow t'_4 = 1 \\ r_0 = -3/4 &\Rightarrow r'_0 = -1/2 \\ r_1 = -1/4 &\Rightarrow r'_1 = -1 + \sqrt{3}/2 \\ r_2 = 1/4 &\Rightarrow r'_2 = 1 - \sqrt{3}/2 \\ r_3 = 3/4 &\Rightarrow r'_3 = 1/2 \end{aligned}$$



$$\begin{aligned} MSE &= E\{(u - u')^2\} \\ &= \sum_{i=0}^3 \int_{t'_i}^{t'_{i+1}} (u - r'_i)^2 p_u(u) du \\ &= \int_{-1}^{-1+\sqrt{2}/2} (u + 1/2)^2 (1+u) du + \int_{-1+\sqrt{2}/2}^0 (u + 1 - \sqrt{3}/2)^2 (1+u) du \\ &\quad + \int_0^{1-\sqrt{2}/2} (u - 1 + \sqrt{3}/2)^2 (1-u) du + \int_{1-\sqrt{2}/2}^1 (u - 1/2)^2 (1-u) du \\ &= 2 \left[\int_{-1}^{-1+\sqrt{2}/2} (u + 1/2)^2 (1+u) du + \int_{-1+\sqrt{2}/2}^0 (u + 1 - \sqrt{3}/2)^2 (1+u) du \right] \\ &= 2(0.0071 + 0.0018) \\ &= 0.0178 \end{aligned}$$

Computation of integral:

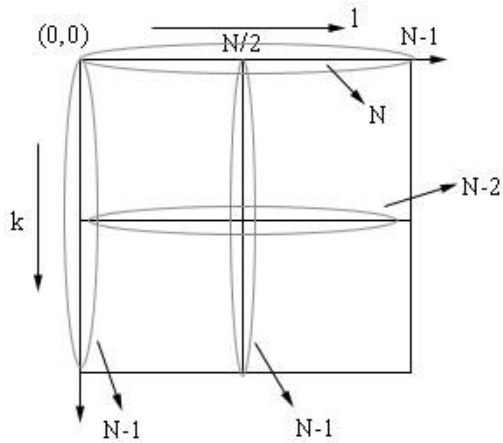
$$\int_a^b (u' + c)^2 (1 + u') du'$$

Let $u = u' + c$, $u' = u - c$, then

$$\int_{a+c}^{b+c} u^2 (1 - c + u) du = \int_{a+c}^{b+c} u^3 du + (1 - c) \int_{a+c}^{b+c} u^2 du$$

$$= 1/4 [(b+c)^4 - (a+c)^4] + (1-c)/3 [(b+c)^3 - (a+c)^3]$$

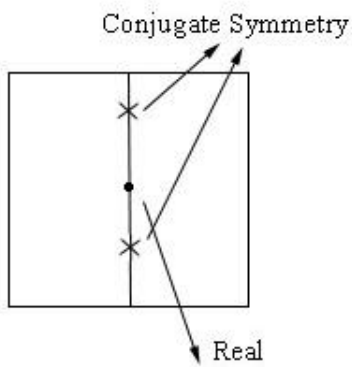
P.3 (34%)



(a) The 1st row is the 1-D DFT of average values of each column. \Rightarrow N real-valued numbers

(b) similar to (a), excluding the origin point \Rightarrow N-1 numbers

(c)



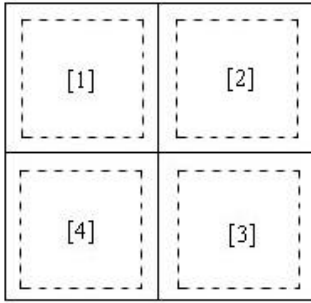
$$v(k, N/2) = v^*(N-k, N/2)$$

$$\therefore \text{Total Number} = (N/2 - 1) \times 2 + 1, \text{ excluding } v(0, N/2)$$

$$= N - 1$$

(d) Similar to (c), Total Number = N-2, excluding $v(N/2, 0)$ and $v(N/2, N/2)$

(e)I



$v(k,l)$ in region [1] is conjugate symmetric with $v(k,l)$ in region [3]

Total number of real numbers

Likewise, there are $= (N/2 - 1)^2 \times 2$ real numbers in regions [2]+[4]

\therefore If we add up the results in (a) – (e)

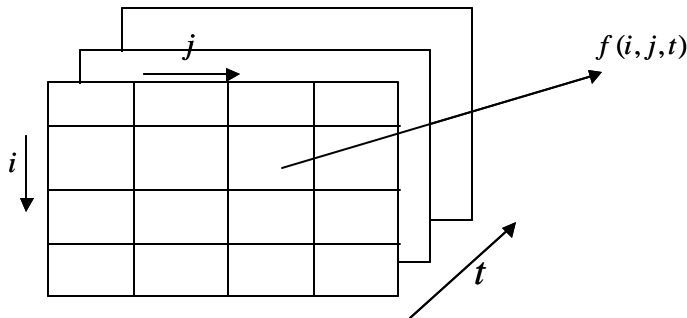
Total number of real numbers

$$= N + (N - 1) + (N - 1) + (N - 2) + (N/2 - 1)^2 \times 2 + (N/2 - 1)^2 \times 2$$

$$= N^2$$

This is the same as the total number of pixels in $u(i, j)$.

P.4. (34%)



Since we are interested in the correlation between features at different time, but the same locations, the input vector will be

$$\mathbf{w}_{i,j} = \begin{bmatrix} f(i, j, t_0) \\ f(i, j, t_1) \\ \dots \\ f(i, j, t_{T-1}) \end{bmatrix}, \text{ where } T \text{ is the time window}$$

From T consecutive image frames, we can get $(N_1/K_1)(N_2/K_2)$ empirical observation of the \mathbf{u} vector.

From the above pool of observation data, we can compute the covariance matrix of \mathbf{u} , $R_{\mathbf{u}}$, which is a T by T matrix.

$$R_{\mathbf{u}}(k, l) = \sum_{i=0}^{N_1/K_1-1} \sum_{j=0}^{N_2/K_2-1} (f(i, j, t_k) - m_k)(f(i, j, t_l) - m_k)$$

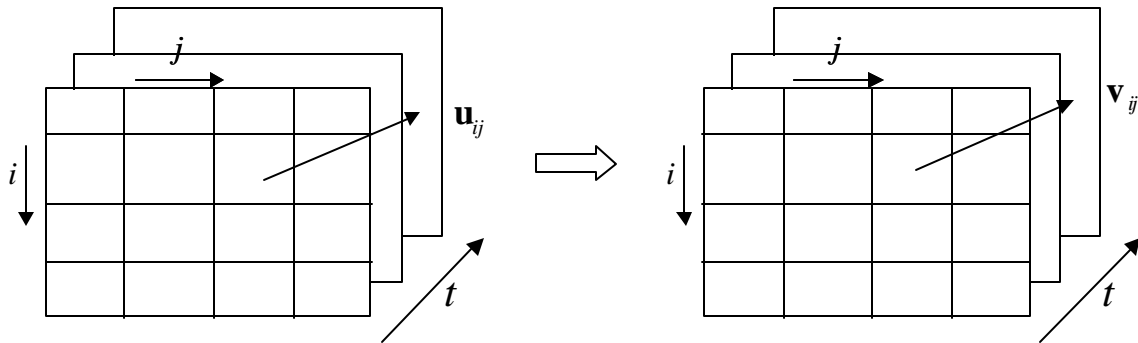
where m_k is the mean of $f(i, j, k)$ over all possible i,j index values.

Find the eigen vectors of $R_{\mathbf{u}}$, $\{\mathbf{f}_0, \mathbf{f}_1, \dots, \mathbf{f}_{T-1}\}$.

Form the KLT matrix as follows

$$\Phi_{T \times T} = [\mathbf{f}_0 | \mathbf{f}_1 | \dots | \mathbf{f}_{T-1}]$$

Now given the original image sequence,



we can use the following KLT transform

$$\mathbf{v}_{ij} = \Phi^{*T} \mathbf{u}_{ij}$$

Here \mathbf{v}_{ij} is a Tx1 vector, Φ is a TxT matrix, and \mathbf{u}_{ij} is a Tx1 vector.

Due to the compaction of energy, some frames of \mathbf{v} near the end can be ignored.