EE5585 Data Compression

March 26, 2013

Lecture 16

Instructor: Arya Mazumdar

1 Review of Homework 6

The review is omitted from this note.

2 Linde-Buzo-Gray (LBG) Algorithm

Let us start with an example of height/weight data.

| H (in) | W (lb) |
|--------|--------|
| 65 | 170 |
| 70 | 170 |
| 56 | 130 |
| 80 | 203 |
| 50 | 153 |
| 76 | 169 |



Figure 1:

We need to find a line such that most points are around this line. This means the distances from the points to the line is very small. Suppose W = 2.5H is the equation of the straight line. Let A be a matrix to project values to this line and its perpendicular direction.

We have

$$A = \begin{bmatrix} 0.37 & 0.92 \\ -0.92 & 0.37 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix}$$
$$X = \begin{bmatrix} 65 \\ 170 \end{bmatrix}, \begin{bmatrix} 70 \\ 170 \end{bmatrix}, \begin{bmatrix} 56 \\ 130 \end{bmatrix} \dots$$

Then by rotating x-axis, we have a new axis:

Y = AX

From the above equation, we can get a set of new H-W table as following:

| H (new) | W (new) |
|---------|---------|
| 182 | 3 |
| 184 | -2 |
| 191 | -4 |
| 218 | 1 |
| 161 | 10 |
| 181 | -9 |

We can get back original data by rotating the axis again: $X = A^{-1}Y$, for example,

$$\left[\begin{array}{c} 65\\170\end{array}\right] = A^{-1} \left[\begin{array}{c} 182\\3\end{array}\right]$$

where,,

$$A^{-1} = \begin{bmatrix} 0.37 & -0.92\\ 0.92 & 0.37 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta)\\ \sin(\theta) & \cos(\theta) \end{bmatrix} = A^T.$$

Note,

$$\begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} * \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

Now, neglect the right column of the new table and set them to 0, that is:

| H (new) | W (new) |
|---------|---------|
| 182 | 0 |
| 184 | 0 |
| 191 | 0 |
| 218 | 0 |
| 161 | 0 |
| 181 | 0 |

Multiplying each row by A^T we have,

| Н | W |
|----|-----|
| 68 | 169 |
| 68 | 171 |
| 53 | 131 |
| 81 | 201 |
| 60 | 151 |
| 67 | 168 |

Compare this table with the original table. We find that the variations are not large. So this new table can be used.

3 Transform Coding

1. For an orthonormal transformation matrix A,

$$A^T = A^{-1} \Leftrightarrow A^T A = A A^T = I. \tag{1}$$

Suppose $y = Ax, \hat{x} = A^{-1}\hat{y}$, where \hat{y} is the stored/compressed value. We introduce error

$$= \|y - \hat{y}\|_{2}^{2} \tag{2}$$

$$= || Ax - A\hat{x} ||_{2}^{2}$$
(3)

$$= || A(x - \hat{x}) ||_{2}^{2}$$
(4)

$$= [A(x - \hat{x})]^{T} * [Ax - \hat{x}]$$
(5)

$$= (x - \hat{x})^{T} * A^{T} * A(x - \hat{x})$$
(6)

$$= (x - \hat{x})^{T} * (x - \hat{x})$$
(7)

$$= \|x - \hat{x}\|_{2}^{2} \tag{8}$$

If we do not introduce a lot of errors in y, then there won't be a lot of error for \mathbf{x} .

2.

$$E[y^T y] = E[x^T A T A x] = E[x^T x].$$
(9)

which means the input and output energies are the same. This is known as Parseval's identity.

4 Hadamard Transform

Suppose,

$$A = 1/\sqrt{2} \begin{bmatrix} 1 & 1\\ 1 & -1 \end{bmatrix}$$
(10)

-1 is shaded as black in the figure below.:





Use Kronecher Product to define: $A_4 = A_2 \oplus A_2$ we have

The Kronecker product is defined to be,

$$\left[\begin{array}{cc}a_{11}&a_{12}\\a_{21}&a_{22}\end{array}\right]\otimes\left[\begin{array}{cc}b_{11}&b_{12}\\b_{21}&b_{22}\end{array}\right]$$

$$= \begin{bmatrix} b_{11}M & b_{12}M \\ b_{21}M & b_{22}M \end{bmatrix}$$
$$M = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

where,

 ${\cal A}_4$ can be represented as below.





Then we can have $A_{16} = A_4 \oplus A_4$ in this similar way.

5 Open Problem: Fix-free code, Conjectures, Update-efficiency

Instructor's note: This section is removed (the reason can be explained in person). Please consult your notes.