MML arrignment 2:Sunday, October 16, 2022 6:33 PM

Tark 1:-

a) Giran data:

•
$$\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$
 $\vec{5}' = \begin{bmatrix} s_1 \\ \vdots \\ s_K \end{bmatrix}$, also $n > K$

whom \vec{x} is the image rector and \vec{z} is the severte mostage.

•
$$\overline{3} = \begin{bmatrix} 3/1 \\ 3/2 \end{bmatrix}$$
 is the modification rector

- We encode the message waing \overline{z} $(\overline{z}+\overline{z})$
- \vec{z} must be signal to insure that original image \vec{x} and $(\vec{x} + \vec{z})$ almost book the same.

• Original maxing in decoded by:
$$D(\vec{z} + \vec{z}) = y$$

$$\hat{S} = nign(y)$$

whose Die a Kxn motrix wied for decoding.

Solution:

Solution:

 \Rightarrow Albor $\vec{3}$ whould be are small are postrible i.e $\vec{3} = \text{arg. min}(|3||)$

This is bost norm budden:

minimize
$$||z||^2$$
 and $|z||^2$ and $|z||^2$ and $|z||^2$ and $|z||^2$ and $|z||^2$ and $|z||^2$ and $|z||^2$

Lograngian function in given by:

$$d(y_g \lambda) = ||y||^2 + \lambda_1 (d_1^T y - e_1) + ... + \lambda_K (d_1^T y - e_K)$$

Here,
$$d_i^T$$
 are now of D

$$\lambda = [\lambda, \dots, \lambda_K]^T$$
 is a lagrange multiplieros

$$\frac{\partial d(3,1)}{\partial \lambda_{i}^{*}} = d_{i}^{T} 3 - e_{i}^{*}$$

$$d_{i}^{T} 3 - e_{i}^{*} = 0 \quad i = 1 \dots K$$

$$d_{i}'y = e_{i} \Rightarrow D^{1}y = E_{i} \rightarrow 0$$

$$\frac{\partial L(y, \Lambda)}{\partial y_{i}} = 2\sum_{i=1}^{n} y_{i} + \sum_{i=1}^{K} \lambda_{i}(d_{i}),$$

$$\frac{\partial L(y, \Lambda)}{\partial y_{i}} = 2y + D^{T}\Lambda$$

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The KKT conditions are:

$$\begin{bmatrix} 2I & D^{T} \end{bmatrix} \begin{bmatrix} \vec{3} \\ \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ \alpha S - D\alpha \end{bmatrix}$$

$$2\vec{3} + \vec{D}^{T} \lambda = 0 \quad \neg \vec{A}$$

$$3\vec{3} = \alpha S - D\alpha \quad \vec{A}$$

$$\vec{J} = \alpha S - D\alpha \quad \vec{A}$$

$$\vec{J} = -\frac{1}{2} \vec{D}^{T} \lambda \rightarrow \vec{C}$$

The property of
$$\Delta M(Q)$$
:

$$-\frac{1}{2}DD^{T}\lambda = \alpha S - Dx$$

$$\lambda = -2(DD^{T})(\alpha S - Dx) - \lambda D$$

$$-\frac{1}{2}D \text{ in a wide matrix, if all the rows we limited indefindent than (DDT) is a innectable.}

Putititing (D) in (Q):$$

Substituting @ im@:

$$2y - 2D^{\dagger}(DD^{\dagger})^{-1}(\alpha S - D\alpha) = 0$$

$$3' = D^{\dagger}(DD^{\dagger})^{-1}(\alpha S - D\alpha)$$

$$3' = D^{\dagger}(\alpha S - D\alpha)$$
where D^{\dagger} is the breador involve of the wide mateux D .

$$y = D^{\dagger}(\alpha S - Dx)$$
where $D^{\dagger} = D^{\dagger}(DD^{\dagger})^{-1}$