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MATH 6397
                                                                     Derive the steps of the ADMM algorithm in Alg L.
                          Step1 x(h+1) (A+A+pF) (ATy+p(Z(h)-u(h))
                           Note that x(12+1) = argmin ly (2,1) = argmin { \frac{1}{2} || Ax-y||_2 + \alpha || \frac{1}{2} || + \frac{1}{2} || x-2+ \alpha || \frac{1}{2} \frac{1}{2}
                                                                                                                                             = argmin 智訓Ax-y||2+专川x-z+u||23。
                          Let A = \begin{bmatrix} a_{11} & a_{11} \\ a_{m1} & a_{mn} \end{bmatrix} \in \mathbb{R}^{m_{1}n}, \chi = \begin{bmatrix} x_{1} \\ y_{1} \end{bmatrix} \in \mathbb{R}^{m_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{1} \end{bmatrix} \in \mathbb{R}^{m_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2} \end{bmatrix} \in \mathbb{R}^{n_{1}}, \chi = \begin{bmatrix} x_{1} \\ y_{2}
                            The partial derivative of la(x,v) wish the considerate the of x is
                                           5/e = 5/2 { = ||Ax - y||_2 + \frac{1}{2} ||x-2+u||_2 }
                                                                         = 5x2 { = (4x) (Ax) - 24 Ax + y y + p (x- =+ w) (x-z+w) }
                                                                      = 最 (主 (是有):本) - 人类的是明知)+ 丁丁十月是(水;-五十七)
                                                                  = = = (ajc(=ajix; -jj)) + f(x==tine).
                               The ATAX - Aty + p(x-z+u). is the gradient of ly (xiv).
                          Set The goal to zero and solve for pe:
                                                          0 = A+Ax - Aty + p (x-2+14)
                                                        Aty + p (2-4) = (ATA + p) x
                                                                                                                                                                                                                                                                                                                                                     Note: ATA+ II invertible for any making A
                                                         x = (ATA+ JI) (ATy+ y (2-4)).
                            5top Z: zik+i) = Suy (xik+)+ u(k))
                               Note that 2(kr) = argain ( (x(en)) will) = argain { a ||2||1 + $\frac{1}{2} \ ||x(\text{lext})||^2 & $\frac{2}{3}$
                                                                                                                                      = argmin { x 清 | = 1 + 支 清 (x; -==; + u; ) = 3.
                                                                                                                                                                                                                                                                                                        NOTE: Subdifferential of tel 6 ( 1), 2000,
                                到=至《《科川+艺名《四五十山》
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        123, 20 > 0.
                                                        = x 5 | = 1 - p (x - = + uc)
                               Find z_c by setting z_c egral to zero; consider all possible values of z_c case: z_c < 0: v_c = -\alpha - p\left(x_c^{(k+1)} - z_c^{(k+1)} + u_c^{(k)}\right)
                                                                                                                                    Z_{c}^{(\hat{p}_{c}+1)} = \chi_{c}^{(\hat{p}_{c}+1)} + \chi_{c}^{(\hat{p}_{c}+1)}
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$$C_{a,5c} = Z_{c} > 0$$
;  $O = A - f(Y_{c}^{(p,1)}) \xrightarrow{(p,1)} + u_{c}^{(p)}$   
 $Z_{c} = X_{c} + u_{c}^{(p)} \xrightarrow{(p,1)} \frac{(p,1)}{(p,1)} \xrightarrow{(p,1)} \frac{(p,1)}{(p,1)} \times u_{c}^{(p)}$ 

$$\frac{3c}{2c^{-2}} = \frac{2c}{2c^{-1}} \frac{1}{1 - p} \left( \frac{4c}{2c} \right) + \frac{1}{2c} \left( \frac{1}{2c} \right) + \frac{1}{2c} \right)$$

$$\frac{(p+1)}{2c} = \frac{1}{2c} \frac{1}{2c} + \frac{1}{2c} \left( \frac{1}{2c} \right) - \frac{1}{2c} \left[ -\frac{1}{2} \cdot \frac{1}{2c} \right]$$

Combrace the derived wordinate wise optimality conditions to obtain 2 (18+1). Extractly, = (ter) = S (x(ter)) + u(ter) = S 1 2 1 (2) = (2) + (1) + (2) + (2) =

Gover with and steps 1 and 2, we obtain when