



实验探究

2018.6.28



$$J = \min_{A, c} \| (X - XA)^T \|_{2,1} + \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1$$

$$\frac{\partial J}{\partial A} = X^T XAU - X^T XU + 4\gamma(AA^T - C)A = 0 \quad U: u_{ii} = \frac{1}{2\|x_i - Xa_i\|_2}$$

For each i compute:

$$u_{ii}$$

For each i compute:

$$a_i = A[:, i]$$

(update M)

update M

compute :

$$\min_c \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1$$

$$u_{ii}X^T Xa_i - u_{ii}X^T x_i + 4\gamma(M - C)a_i = 0$$

$$M = AA^T$$

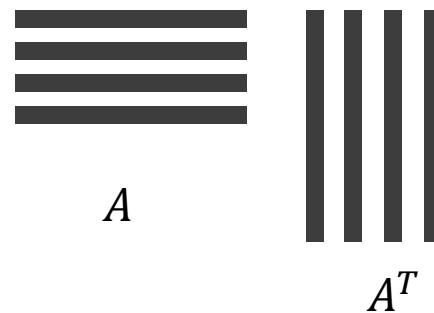
$$a_i = u_{ii}[u_{ii}X^T X + 4\gamma(M - C)]^{-1}X^T x_i$$

$$M = AA^T A \quad u_{ii}X^T Xa_i - u_{ii}X^T x_i + 4\gamma m_i - 4\gamma C a_i = 0$$

$$a_i = [u_{ii}X^T X - 4\gamma C]^{-1}[u_{ii}X^T x_i - 4\gamma m_i]$$



$$\min_{A,c} \gamma \|AA^T - C\|_F^2 + \lambda \|c\|_1$$



对于式(11.12), 可先计算 $z = x_k - \frac{1}{L} \nabla f(x_k)$, 然后求解

$$x_{k+1} = \arg \min_x \frac{L}{2} \|x - z\|_2^2 + \lambda \|x\|_1. \quad (11.13)$$

令 x^i 表示 x 的第 i 个分量, 将式(11.13)按分量展开可看出, 其中不存在 $x^i x^j$ ($i \neq j$) 这样的项, 即 x 的各分量互不影响, 于是式(11.13)有闭式解

$$x_{k+1}^i = \begin{cases} z^i - \lambda/L, & \lambda/L < z^i; \\ 0, & |z^i| \leq \lambda/L; \\ z^i + \lambda/L, & z^i < -\lambda/L, \end{cases} \quad (11.14)$$

其中 x_{k+1}^i 与 z^i 分别是 x_{k+1} 与 z 的第 i 个分量. 因此, 通过 PGD 能使 LASSO 和其他基于 L_1 范数最小化的方法得以快速求解.

1. Fix the others and update J by setting $J = \arg \min_J \frac{1}{\mu} \|J\|_* + \frac{1}{2} \|J - (Z + Y_2/\mu)\|_F^2$.
2. Fix the others and update S by setting $S = \arg \min_S \frac{1}{\mu} \|S\|_* + \frac{1}{2} \|S - (L + Y_3/\mu)\|_F^2$.
3. Fix the others and update Z by setting $Z = (I + X^T X)^{-1} (X^T (X - LX - E) + J + (X^T Y_1 - Y_2)/\mu)$.
4. Fix the others and update L by setting $L = ((X - XZ - E)X^T + S + (Y_1 X^T - Y_3)/\mu)(I + X X^T)^{-1}$.
5. Fix the others and update E by setting $E = \arg \min_E \lambda/\mu \|E\|_1 + 0.5 \|E - (X - XZ - LX + Y_1)/\mu\|_F^2$.
6. Update the multipliers by $Y_1 = Y_1 + \mu(X - XZ - LX - E)$, $Y_2 = Y_2 + \mu(Z - J)$, $Y_3 = Y_3 + \mu(L - S)$.

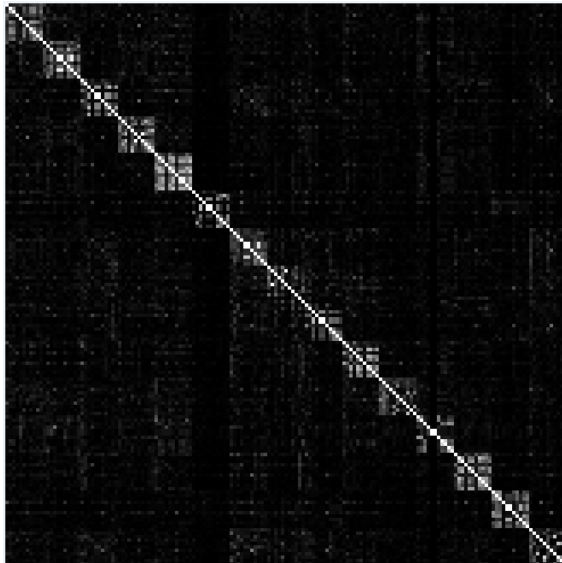
$$\frac{\partial J}{\partial A} = X^T XAU - X^T XU + 4\gamma(AA^T - C)A = 0$$

$$\gamma = 0 \rightarrow A = I$$

$$J = \min_{A,c} \|(X - XA)^T\|_{2,1} + \gamma\|AA^T - C\|_F^2 + \lambda\|c\|_1 + \beta\text{Tr}(A)$$

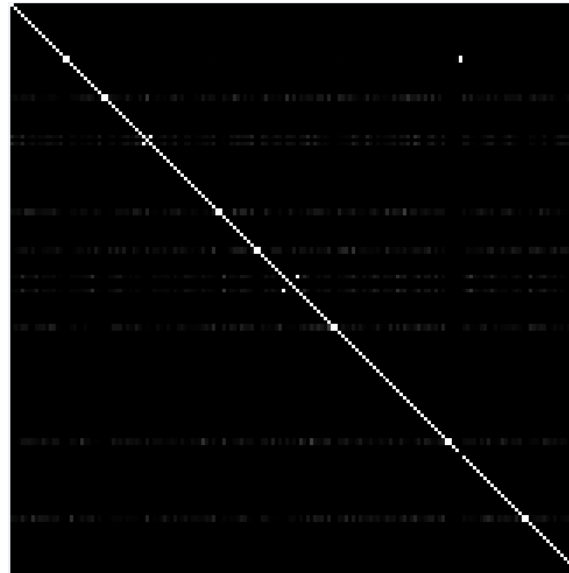
$$J = \min_A \|(X - XA)^T\|_{2,1} + \gamma\|A\|_{2,1}$$

$$\gamma = 20$$

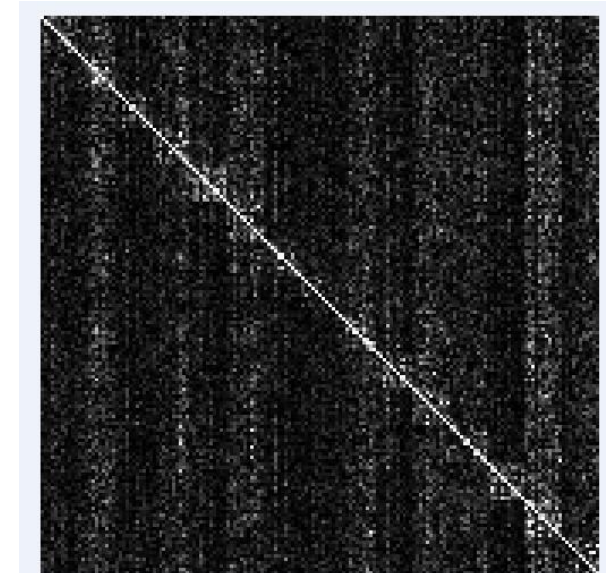


$$J = \min_{A,c} \|(X - XA)^T\|_{2,1} + \gamma\|AA^T - C\|_F^2 + \lambda\|c\|_1$$

$$\gamma = 0.1$$



$$\gamma = 25$$





$$J = \min_{A, c} \| (X - XA)^T \|_{2,1} + \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1 + \beta \text{Tr}(A)$$

For each i compute:

$$u_{ii}$$

For each i compute:

$$a_i = A[:, i]$$

(update M)

compute :

$$\min_c \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1$$

$$U: u_{ii} = \frac{1}{2 \| x_i - Xa_i \|_2}$$

$$\frac{\partial J}{\partial A} = X^T XAU - X^T XU + 4\gamma(AA^T - C)A + \beta I = 0$$

$$M = AA^T$$

$$u_{ii}X^T Xa_i - u_{ii}X^T x_i + 4\gamma(M - C)a_i + \beta I = 0$$

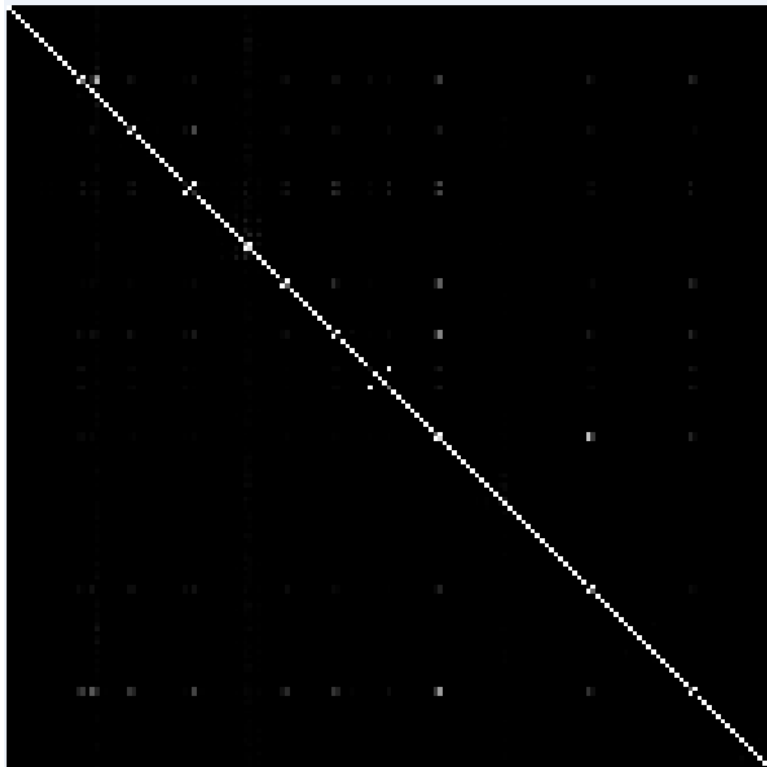
$$a_i = [u_{ii}X^T X + 4\gamma(M - C)]^{-1} [u_{ii}X^T x_i - \beta I_i]$$

$$\text{condition} = \| A^{(k+1)} - A^{(k)} \|_\infty$$

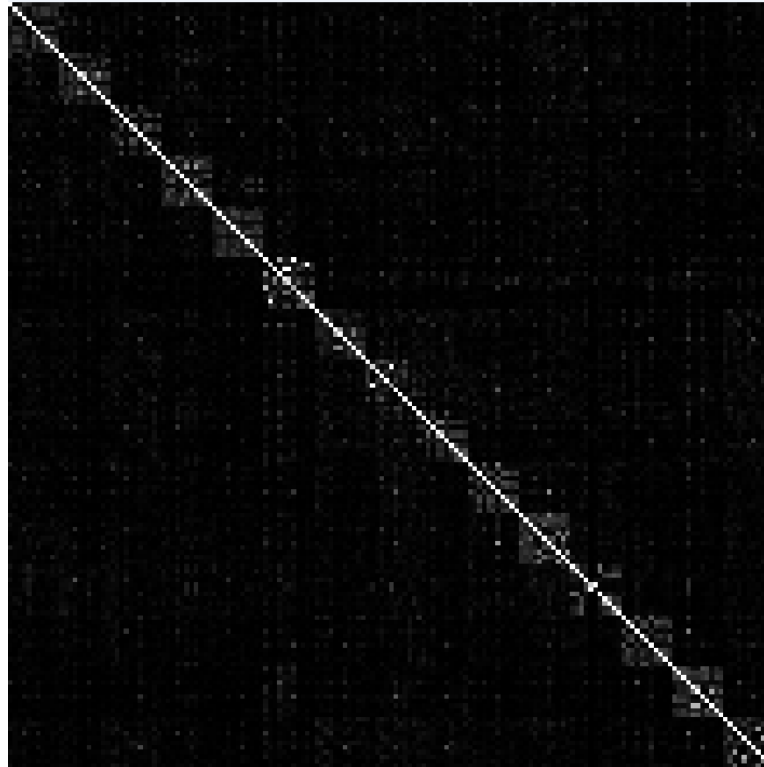
$$J = \min_{A,c} \|(X - XA)^T\|_{2,1} + \gamma \|AA^T - C\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

$$\gamma = 0.1 \quad \lambda = 100$$

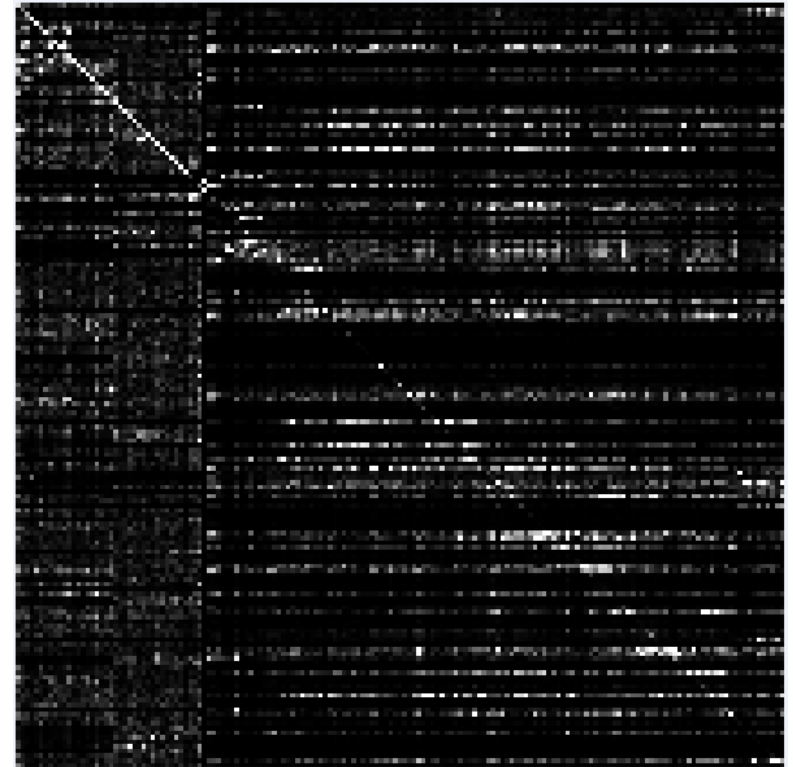
$$\beta = 1$$



$$\beta = 10$$



$$\beta = 25$$



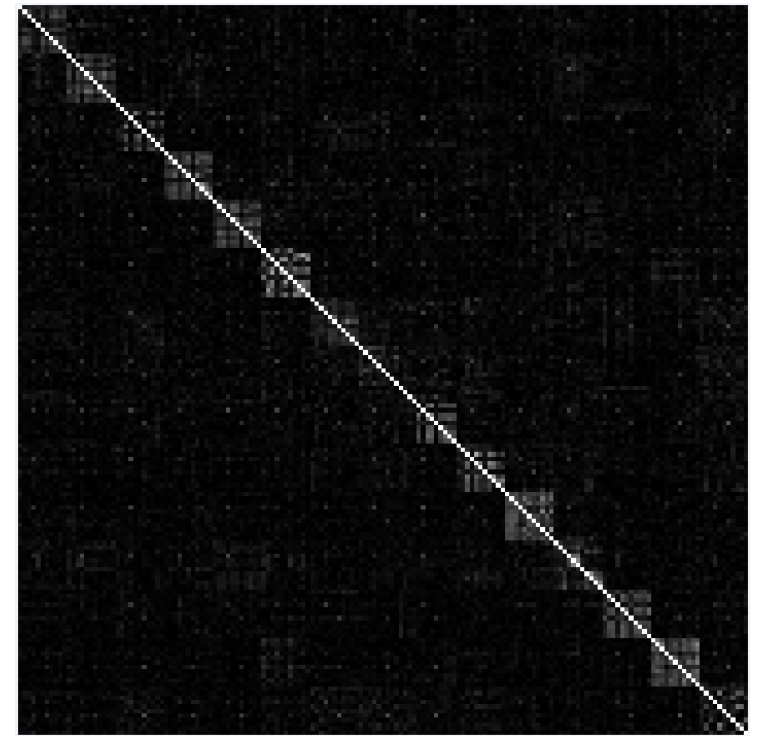
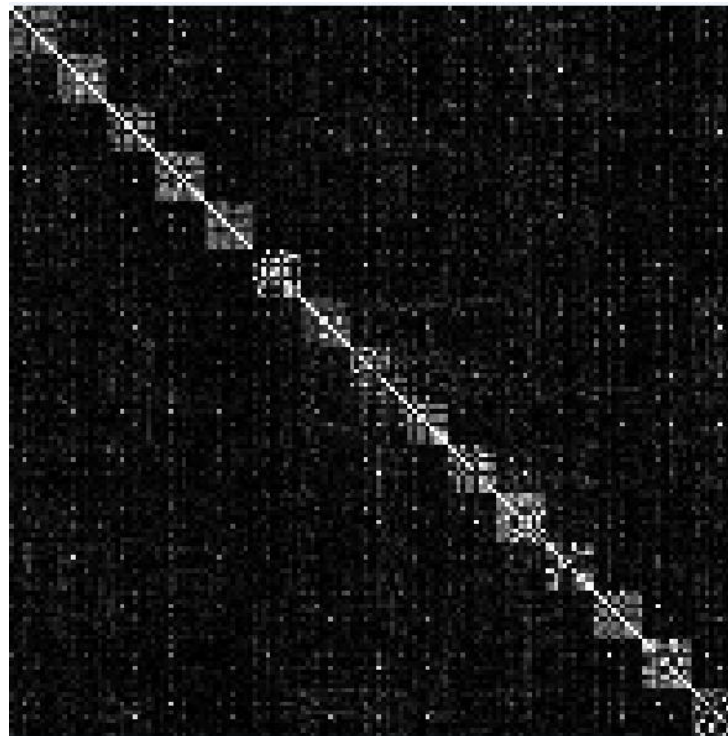
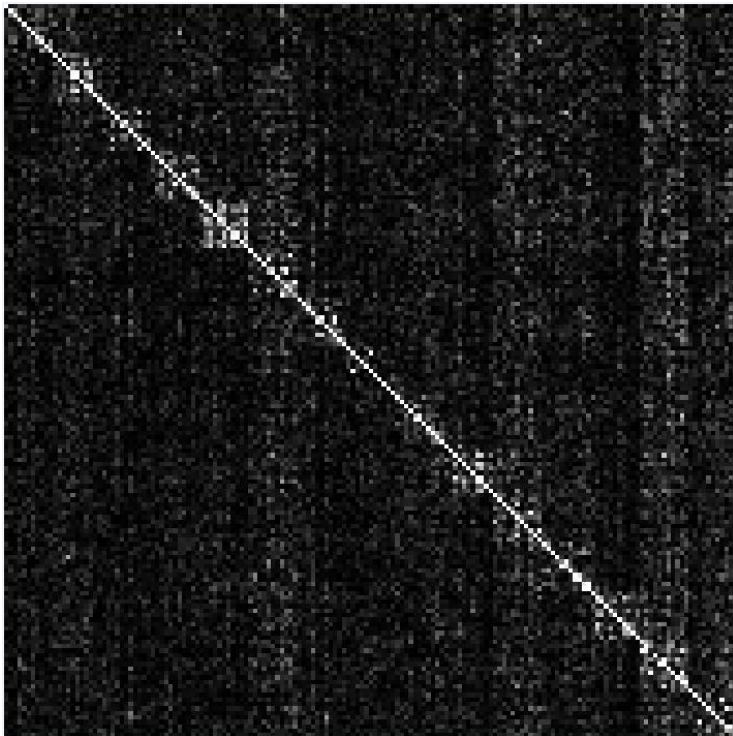
$$J = \min_{A,c} \|(X - XA)^T\|_{2,1} + \gamma \|AA^T - C\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

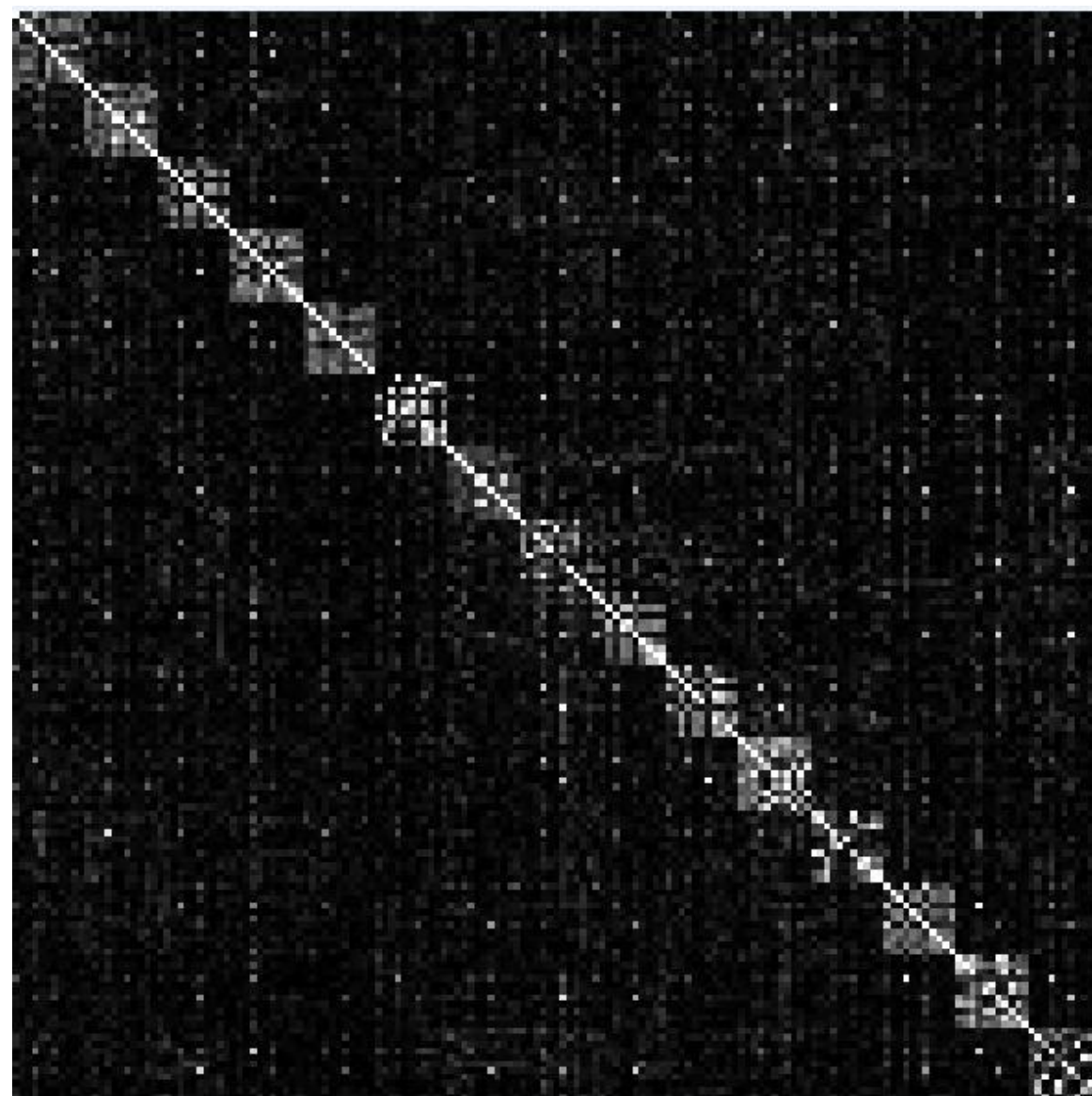
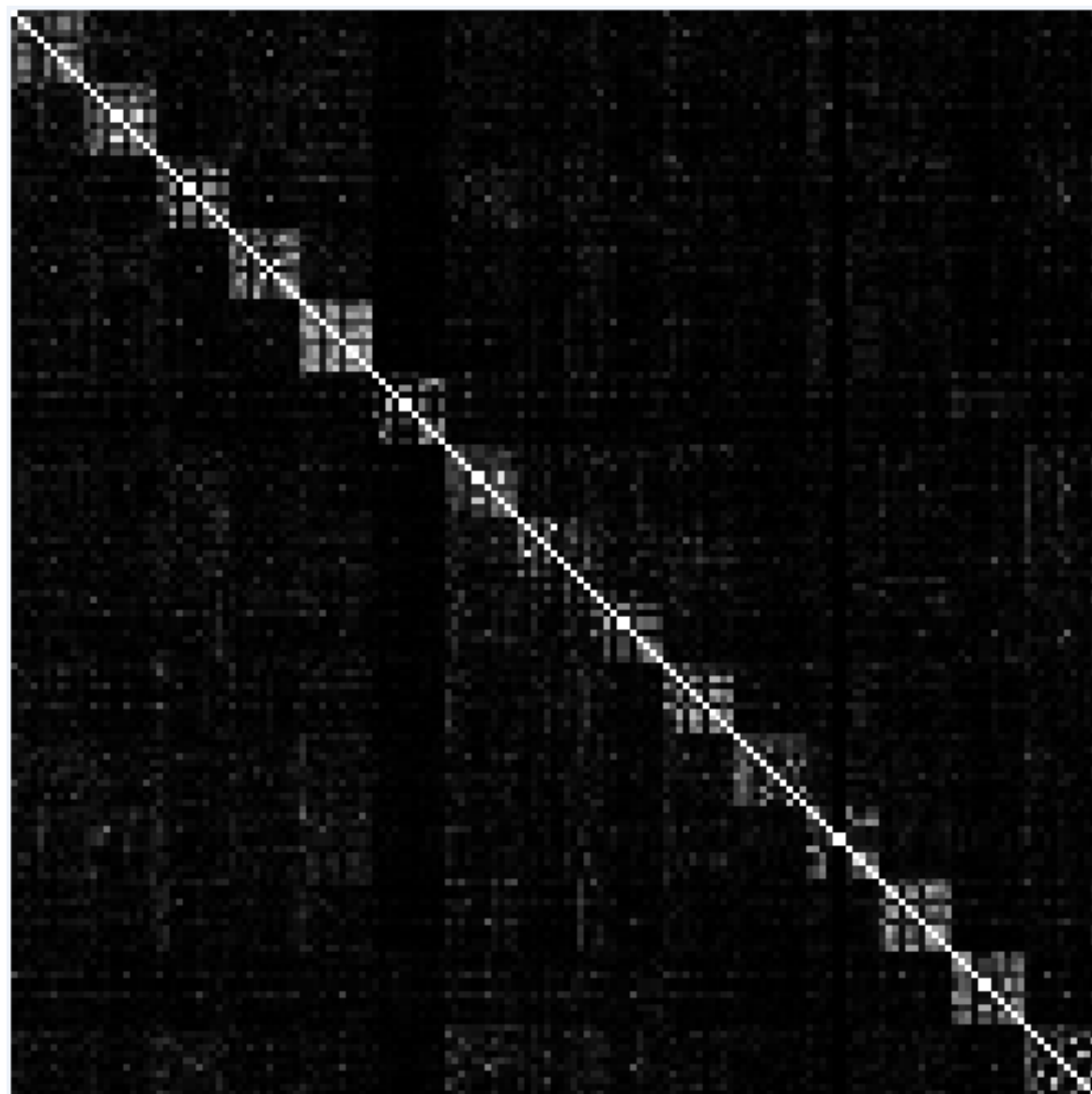
$$\gamma = 20 \quad \lambda = 1000$$

$$\beta = 1$$

$$\beta = 10$$

$$\beta = 25$$







$$J = \min_{A, c} \| (X - XA)^T \|_{2,1} + \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1 + \beta \text{Tr}(A)$$

$$\min_c \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1$$

$$J = \min_{A, c} \| (X - XA)^T \|_{2,1} + \gamma \| \alpha I \circ AA^T - C \|_F^2 + \lambda \| c \|_1 + \beta \text{Tr}(A)$$

$$\min_c \alpha \gamma \| AA^T - C \|_F^2 + \lambda \| c \|_1$$

$$M = AA^T$$

$$u_{ii} X^T X a_i - u_{ii} X^T x_i + 4\gamma (\alpha I \odot M - C) a_i + \beta I = 0$$

$$a_i = [u_{ii} X^T X + 4\gamma (\alpha I \odot M - C)]^{-1} [u_{ii} X^T x_i - \beta I_i]$$



发散

0 132.754615871
1 250.022028446
2 246.974418147
3 67.1678533437
4 3895.17266484
5 2.60201592225e+13
6 3.38989496042e+15
7 6.27534517879e+15
8 3.40474522527e+15
9 5.69296228877e+14
10 7.08299711849e+13
11 5.61912608559e+13
12 1.10123962491e+14
13 1.2769774467e+14
14 1.25850866227e+14

不稳定收敛

0 298.306167738
1 249.396290217
2 63.2561886152
3 28.0695339225
4 13.4056337035
5 8.03934969875
6 7.16633535709
7 8.01880042924
8 10.5997939241
9 9.68991452295
10 11.3607668235
11 10.3851267168
12 8.32750860286
13 9.57717915154
14 17.3764564148

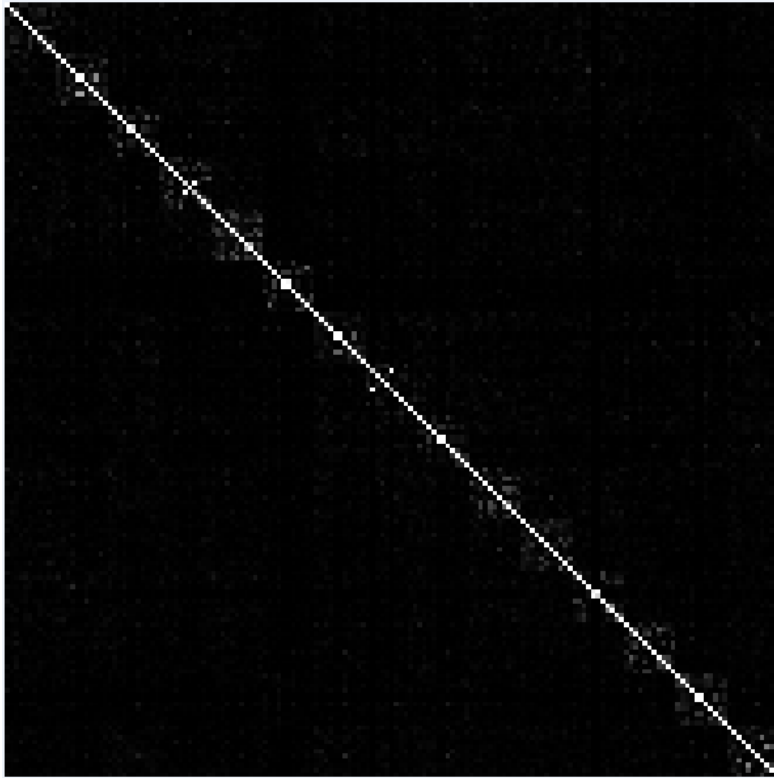
稳定收敛

0 91.6665835061
1 2.18943673032
2 1.71088645991
3 1.63700719974
4 0.806797026423
5 0.5676192167
6 0.410936619691
7 0.164639672102
8 0.0958034126188
9 0.0370229018337
10 0.0146193499335
11 0.00710508556323
12 0.0030408995484
13 0.00141397072025
14 0.000642247208244
15 0.000282608451969
16 0.000136134408896
17 5.66143035831e-05
18 2.88865602959e-05
19 1.13433723171e-05

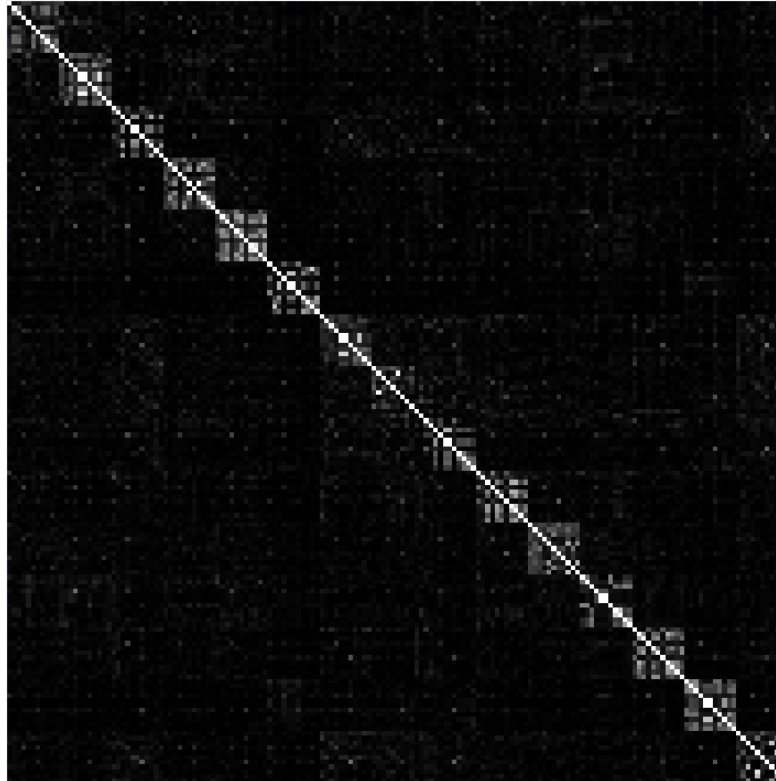
$$J = \min_{A,c} \|(X - XA)^T\|_{2,1} + \gamma \|\alpha I \circ AA^T - C\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

$$\gamma = 0.5 \quad \lambda = 100 \quad \beta = 0$$

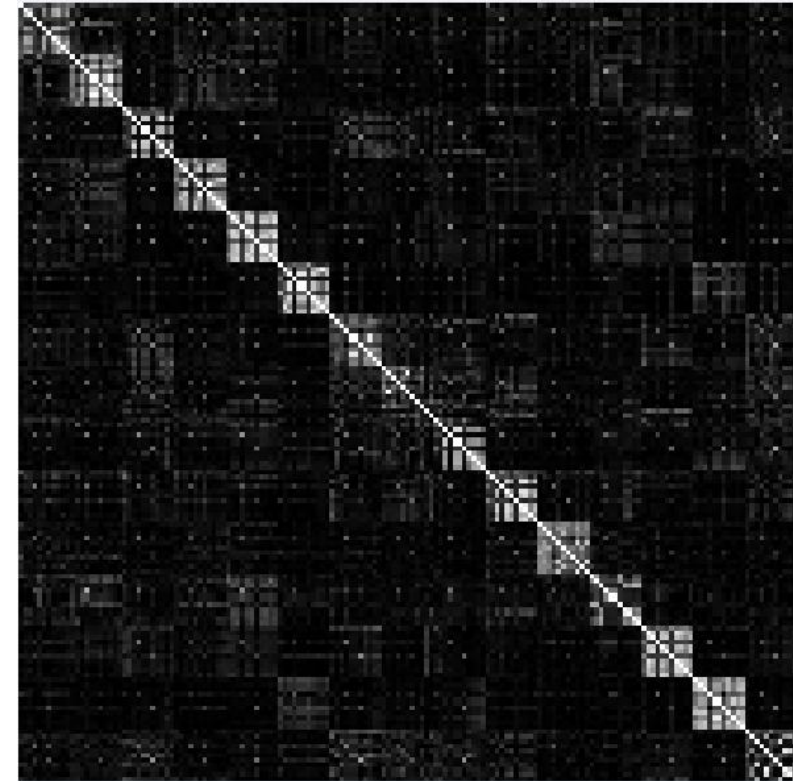
$\alpha = 10$



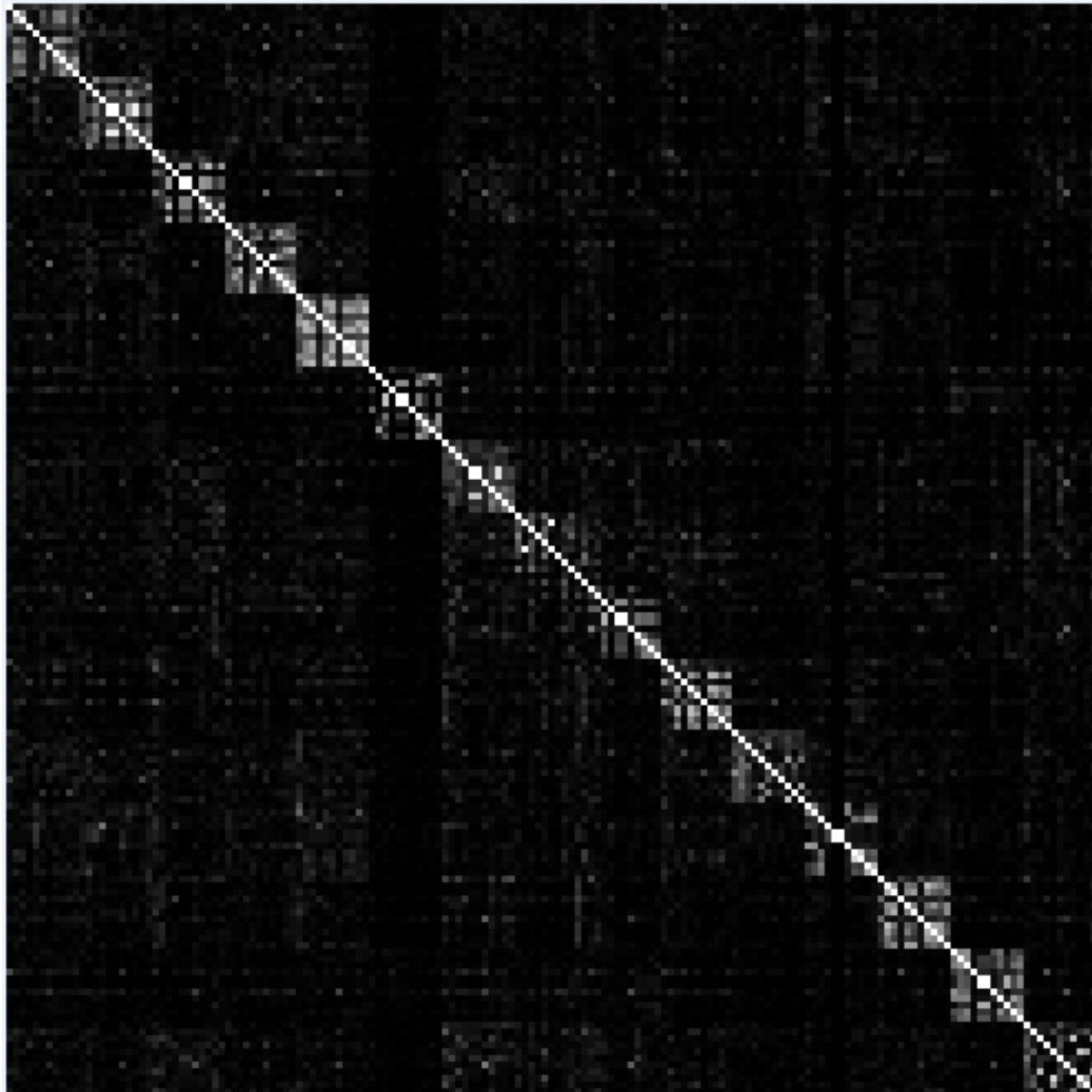
$\alpha = 100$



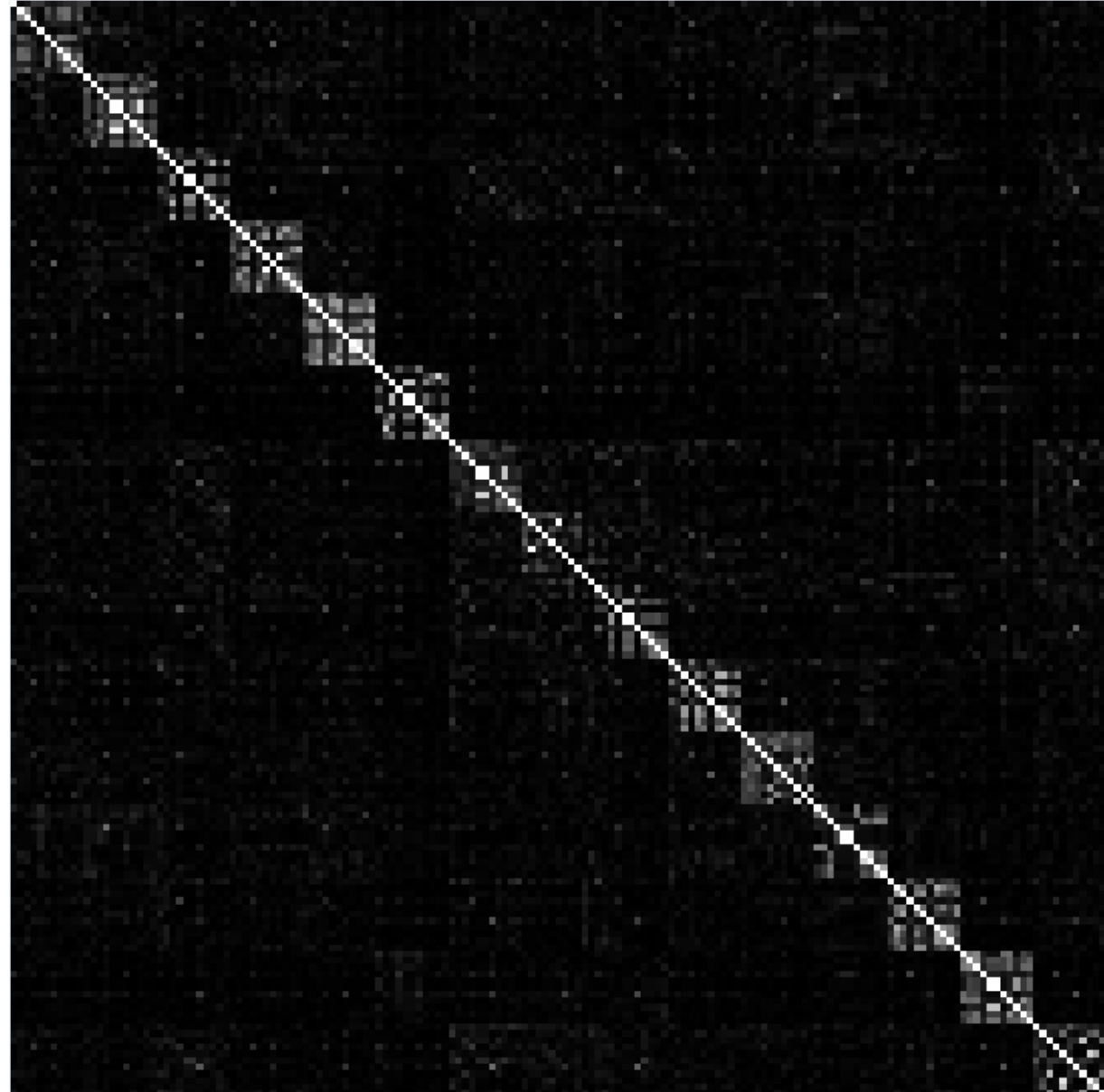
$\alpha = 1000$



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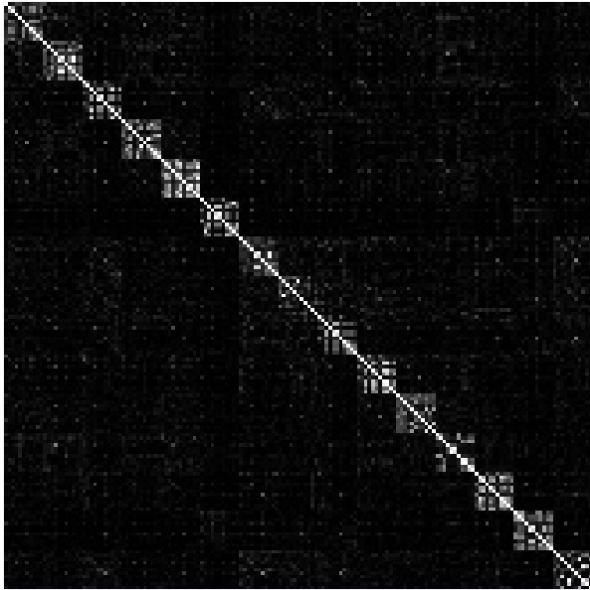
$\alpha = 100$



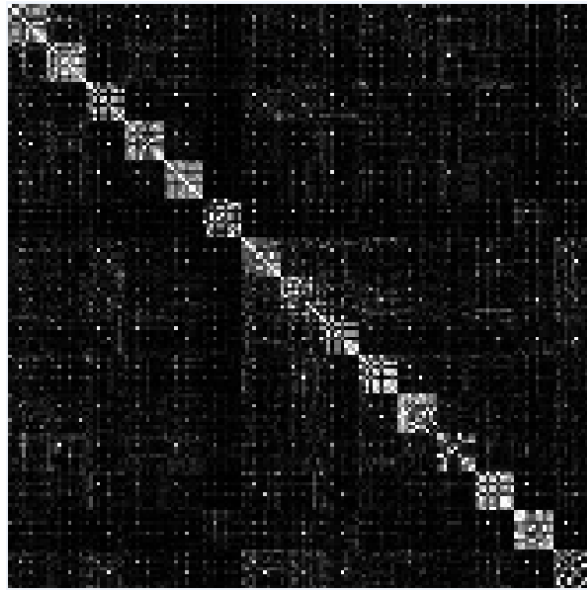
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$$\gamma = 0.5 \quad \lambda = 100 \quad \alpha = 100$$

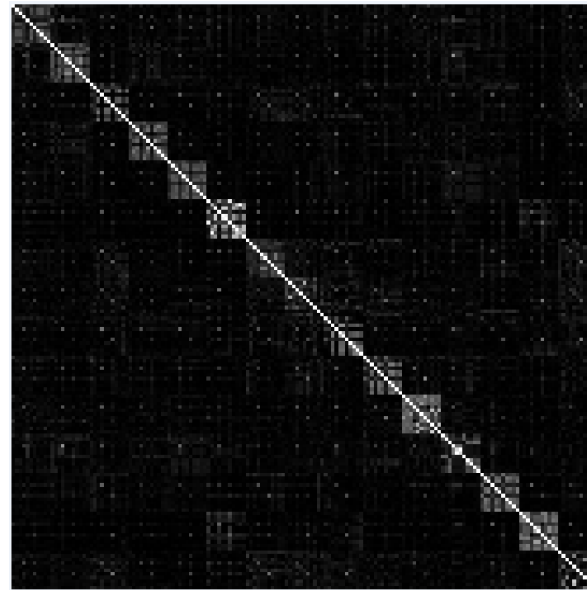
$$\beta = 1$$



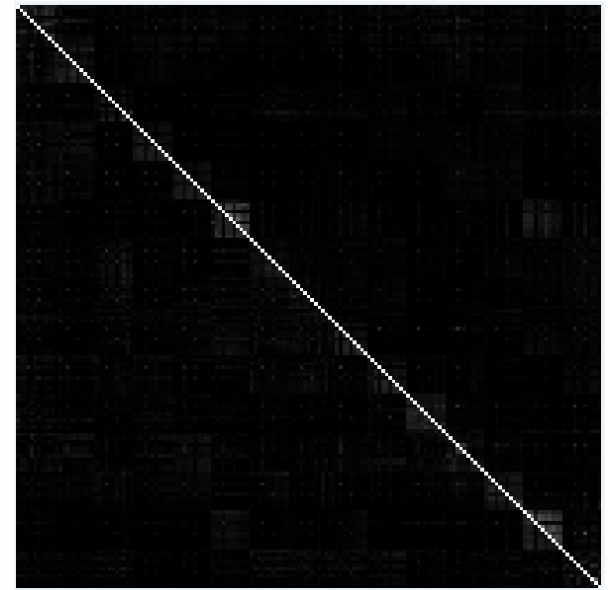
$$\beta = 10$$



$$\beta = 20$$



$$\beta = 100$$



$$J = \min_{A, c} \left\| (X - XA)^T \right\|_{2,1} + \gamma \left\| \alpha I \circ AA^T - C \right\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

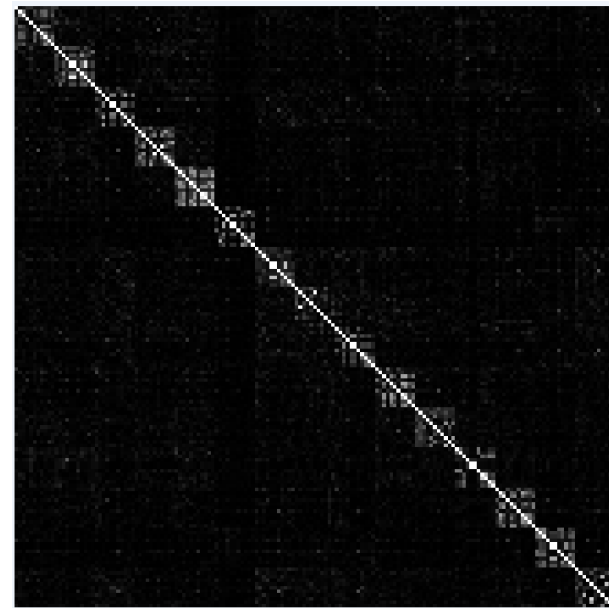
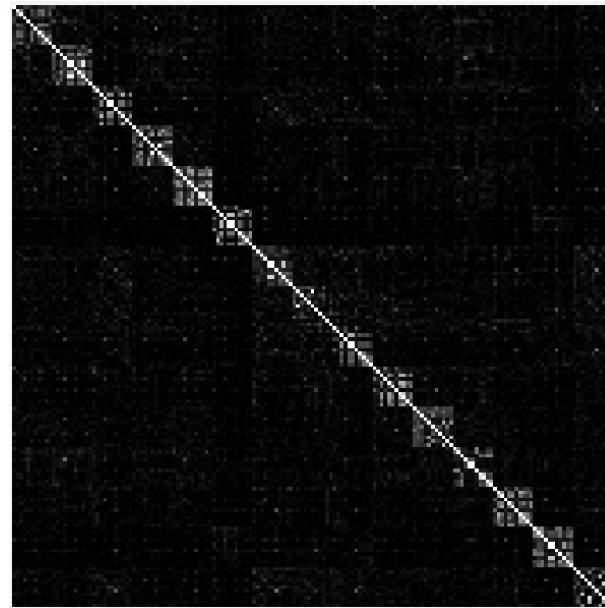
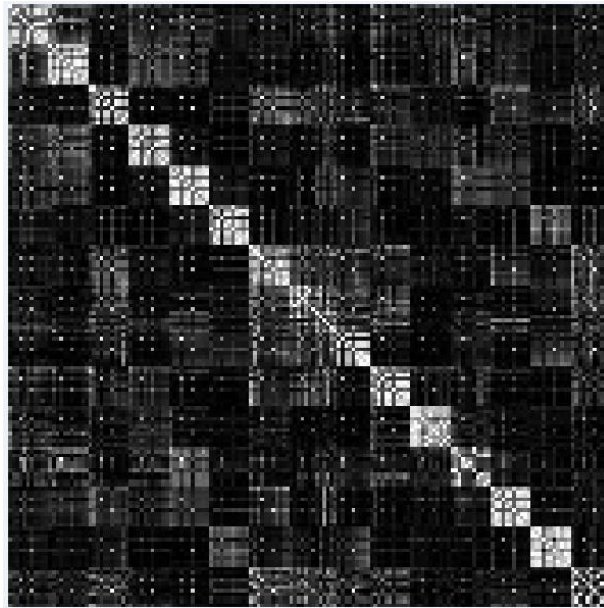
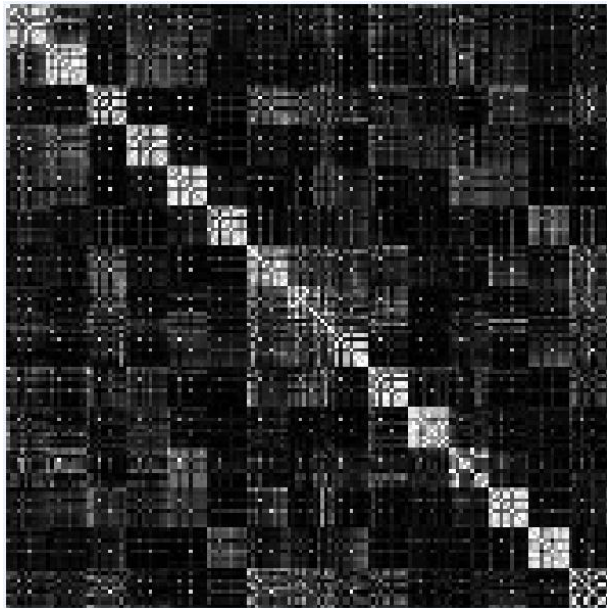
$$\alpha = 100$$

$$\gamma = 20 \quad \lambda = 100 \quad \beta = 4$$

$$\gamma = 20 \quad \lambda = 1000 \quad \beta = 4$$

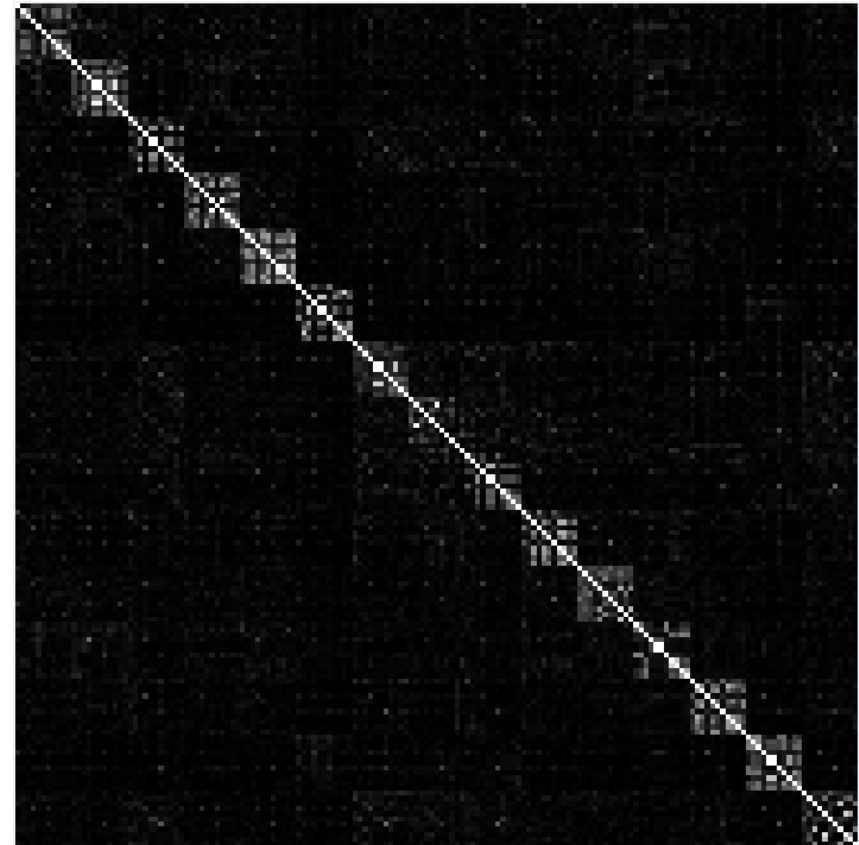
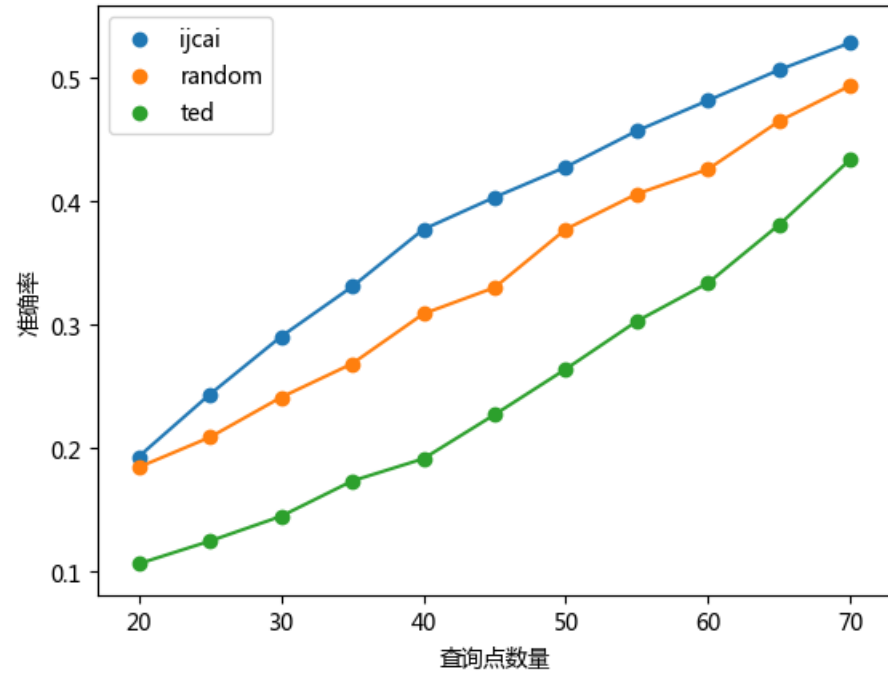
$$\gamma = 0.5 \quad \lambda = 100(1000) \quad \beta = 0$$

$$\gamma = 0.5 \quad \lambda = 10 \quad \beta = 0$$



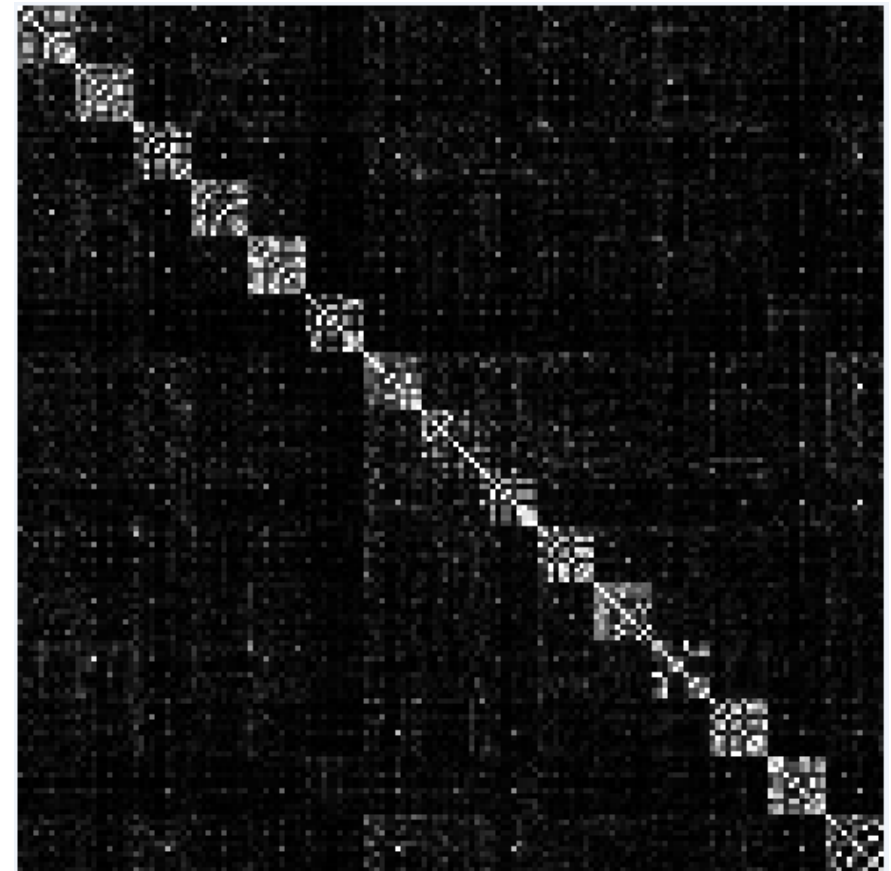
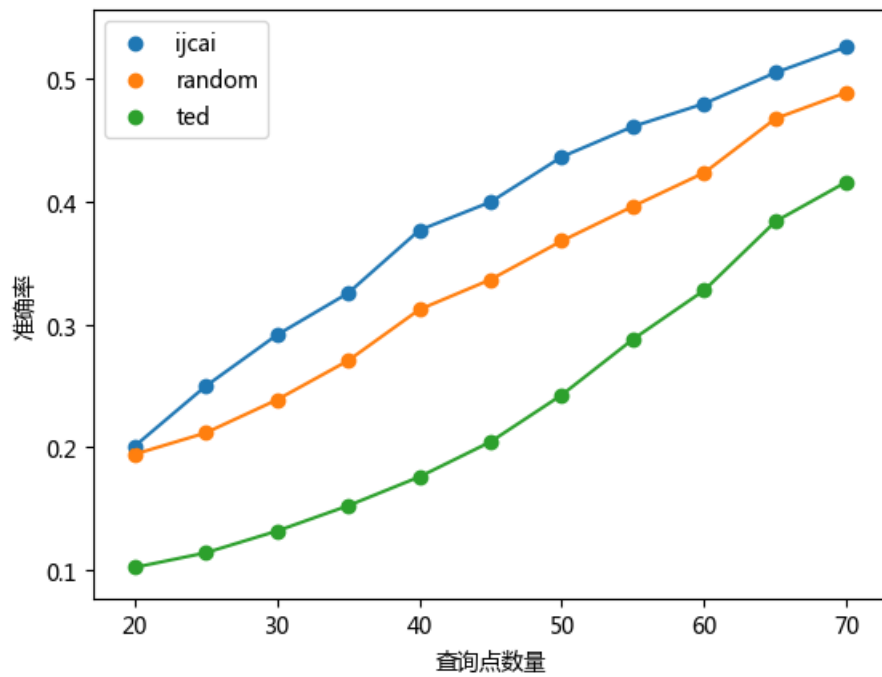
$$J = \min_{A, c} \|(X - XA)^T\|_{2,1} + \gamma \|\alpha I \circ AA^T - C\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

$$\gamma = 0.5 \quad \lambda = 100 \quad \alpha = 100 \quad \beta = 0$$

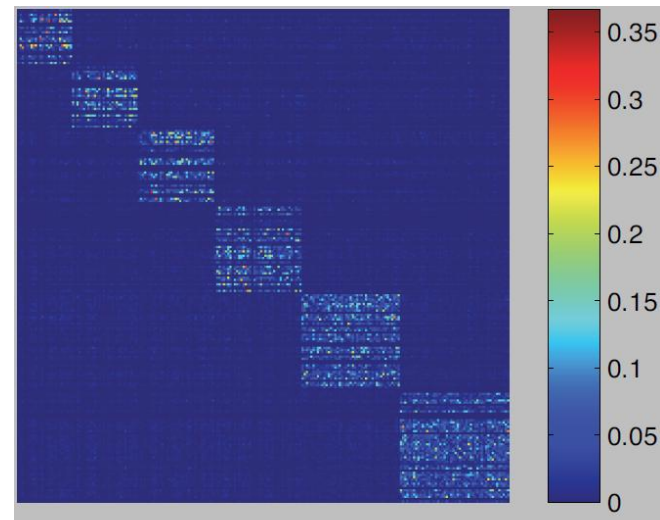
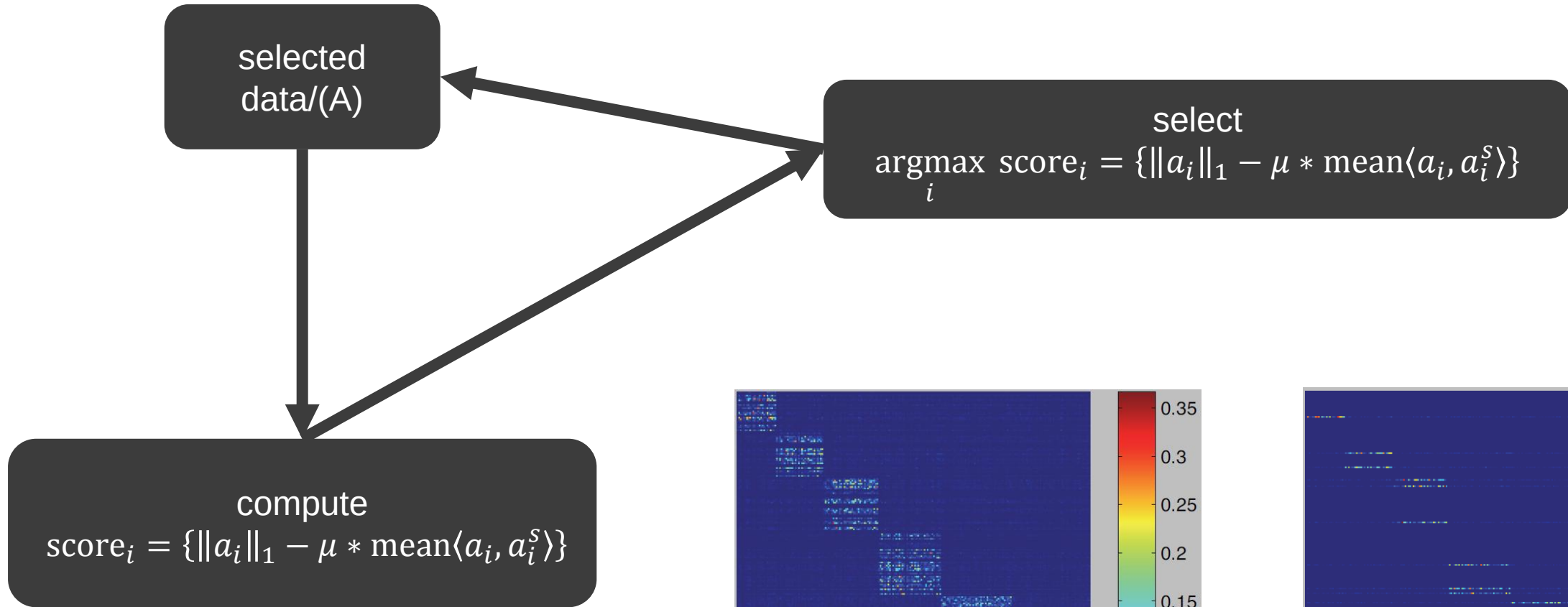


$$J = \min_{A, c} \left\| (X - XA)^T \right\|_{2,1} + \gamma \left\| \alpha I \circ AA^T - C \right\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

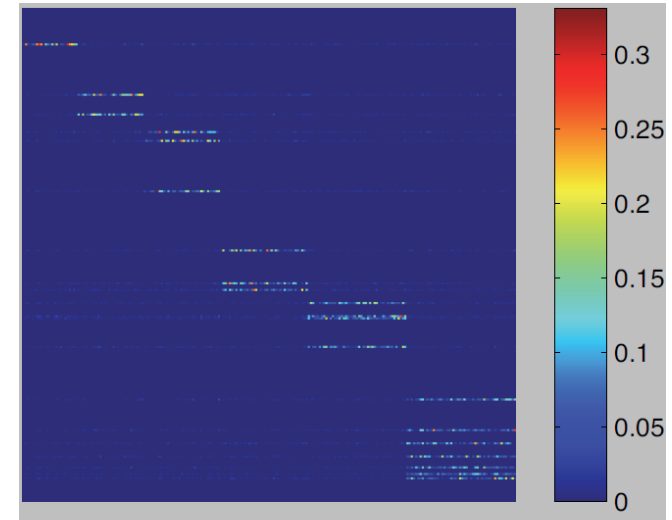
$$\gamma = 0.5 \quad \lambda = 100 \quad \alpha = 100 \quad \beta = 4$$



a_i^s :已经选中的数据点对应的A的所在行



(b) Learned matrix A



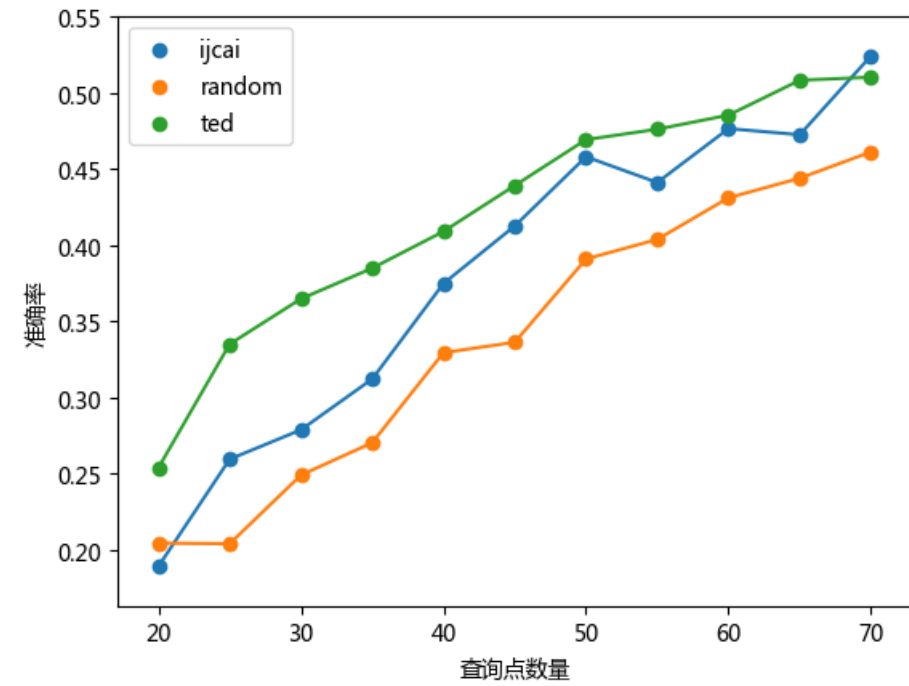
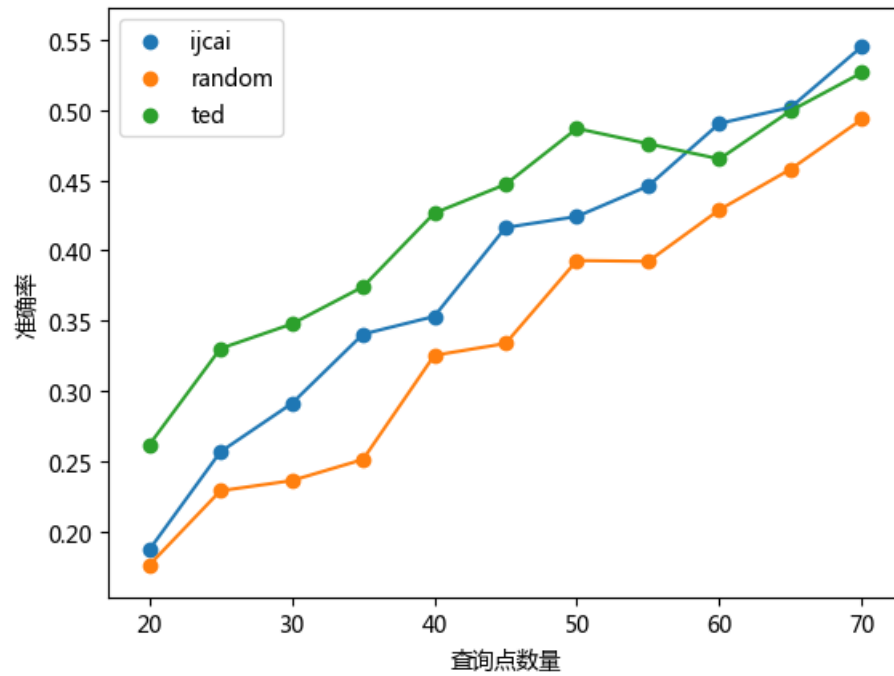
(c) The top 20 rows of A



$$J = \min_{A,c} \|(X - XA)^T\|_{2,1} + \gamma \|\alpha I \circ AA^T - C\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

yaleface 11*15=155 traindata: 77 testdata: 78

$\gamma = 0.5$ $\lambda = 100$ $\alpha = 100$ $\beta = 4$





$$J = \min_{A, c} \|(X - XA)^T\|_{2,1} + \gamma \|\alpha I \odot AA^T - C\|_F^2 + \lambda \|c\|_1 + \beta \text{Tr}(A)$$

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