

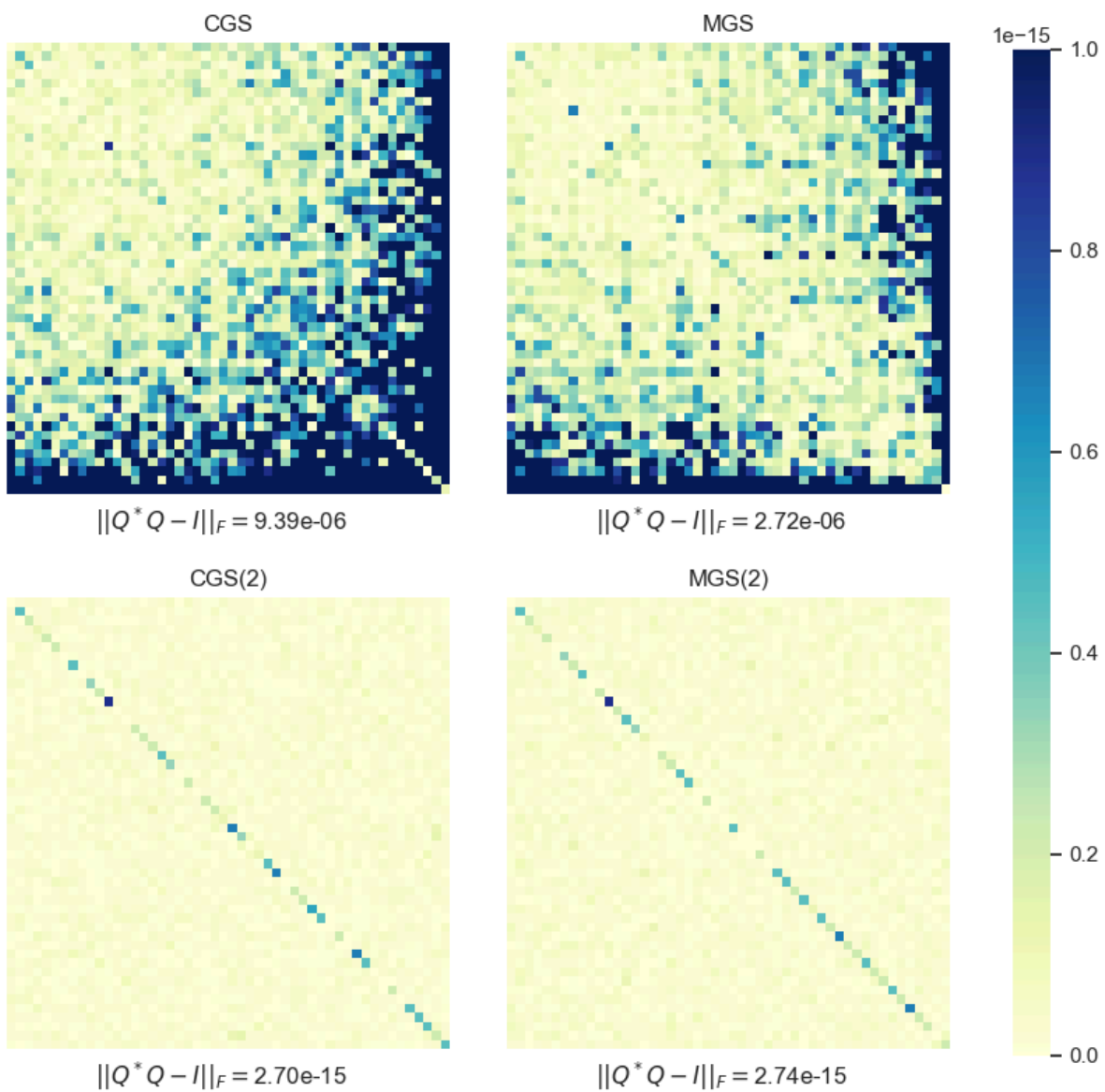
# Homework 5 Solutions

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## Problem 1

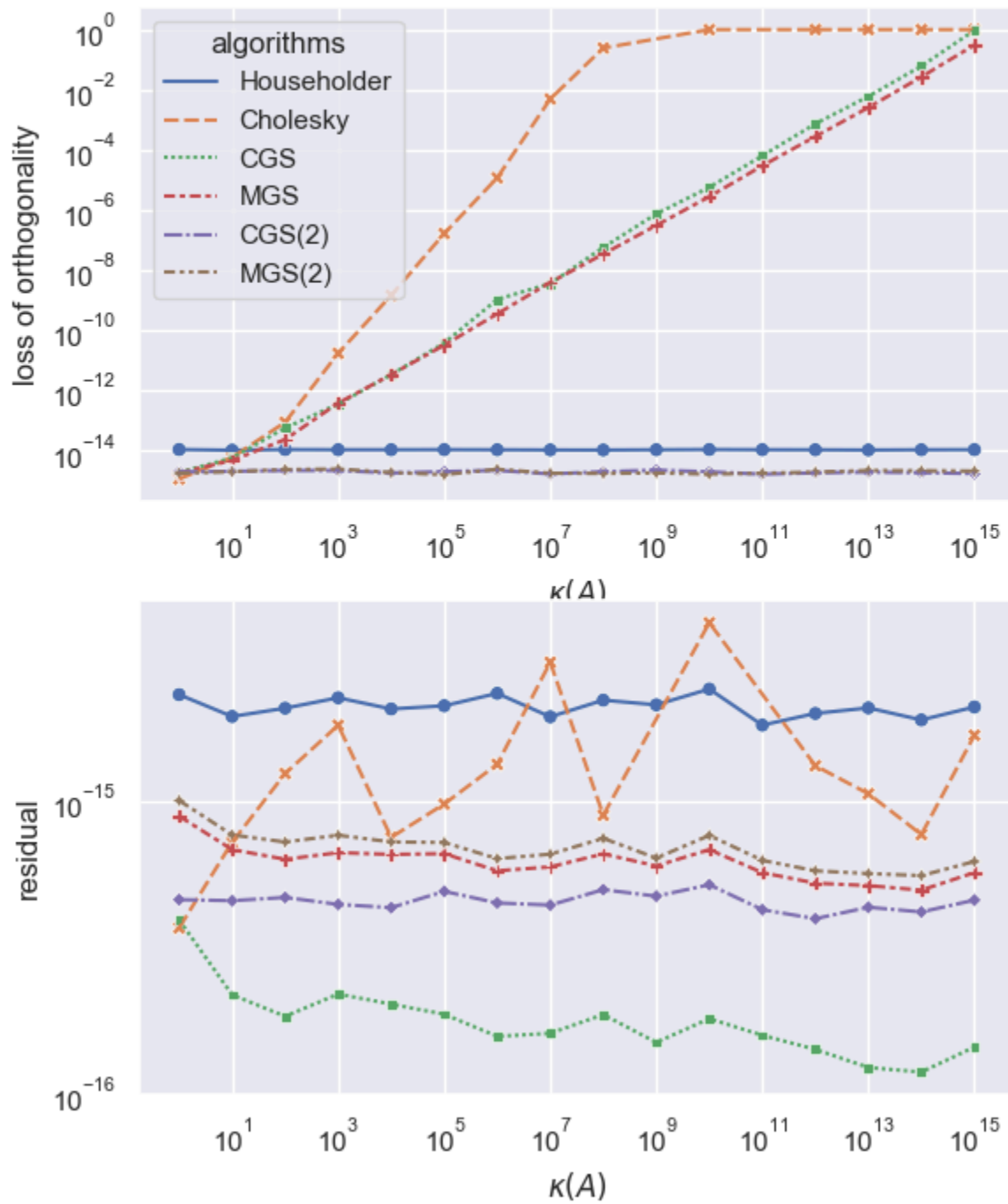
See `qrdecomp.py` .

Orthogonal Loss of Gram-Schmidt QR (size =  $200 \cdot 50$ ,  $\kappa = 10^{10}$ )



## Problem 2

See `qrdecomp.py` .



### Problem 3

For all minimizer  $x$  of the least square problem  $\min ||Ax - b||_2$ ,  $x$  must satisfy

$$A^*Ax = A^*b.$$

Hence for any  $b \in \mathbb{C}^m$ ,

$$A^*AXb = A^*b.$$

Let  $b$  run through standard basis in  $\mathbb{C}^m$ , we see that every column of  $A^*AX$  and  $A^*$  is the same, hence

$$A^*AX = A^*, \tag{3.1}$$

then we also have

$$\begin{aligned} (AX)^*A &= A, \\ X^*A^*AX &= X^*A^*, \end{aligned}$$

combine results above, we have

$$AX = (AX)^*AX = X^*A^*AX = X^*A^* = (AX)^*,$$

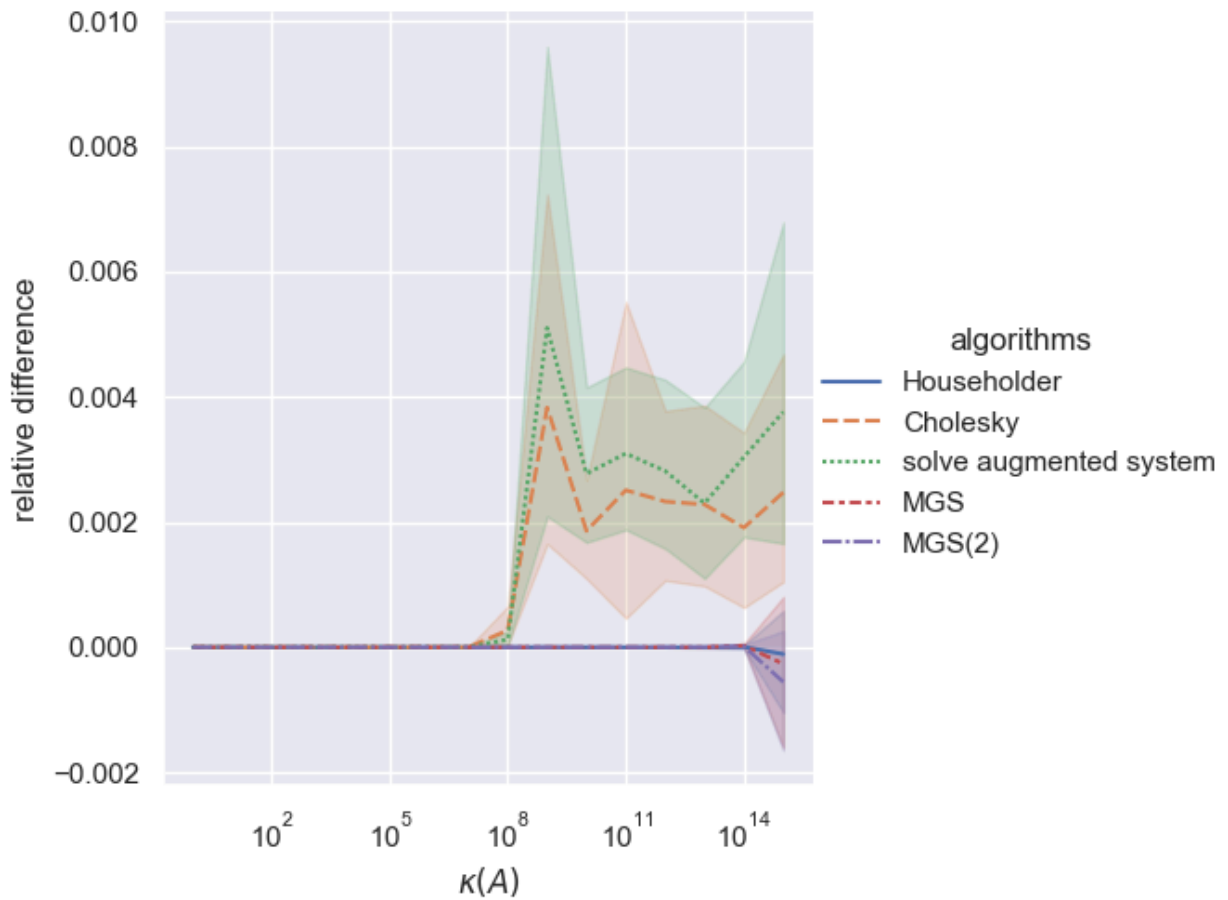
and

$$AXA = (AX)^*A = A. \quad \square$$

### Problem 4

See `lstsq.py`.

For a residual  $r$  from given solution, the "relative difference" means  $\frac{r-r_0}{r_0}$ , in which  $r_0$  is the residual from the solution given by `np.linalg.lstsq`.



### Problem 5

Let the approximate straight line be  $\hat{y} = a_1x + a_2$ , and we use

$$\min \sum_{i=1}^6 (y - \hat{y})^2$$

to determine whether the solution is best. This can be easily transformed in to a least square problem  $\min \|Ax - b\|_2$ , in which

$$A = \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ x_3 & 1 \\ x_4 & 1 \\ x_5 & 1 \\ x_6 & 1 \end{bmatrix}, b = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \\ y_6 \end{bmatrix}.$$

For the solution and visualization, see `lstsq.py`.

