

Problem 1

Since A is symmetric and has k distinct eigenvalues, the minimal polynomial of A has order k , then $\forall r_0 \in \mathbb{R}^n, m \geq k + 1$, the Krylov subspace $\mathcal{K}(A, r_0, m) = \mathcal{K}(A, r_0, k + 1)$.

Then after k iterates of CG, x_{k+1} minimize $\|x - x_*\|_A$ in $\mathcal{K}(A, r_0, k + 1) = \mathcal{K}(A, r_0, n) = \mathbb{R}^n$, hence converges.

Problem 2

Suppose we are going to use PCG on $Ax = b$, and the preconditioner (also symmetric and positive definite) is M . Let C be the square root of M .

Let

$$\begin{aligned}\tilde{A} &= C^{-1}AC^{-1}, \\ \tilde{x} &= Cx, \\ \tilde{b} &= C^{-1}b,\end{aligned}$$

and we write the scheme of CG on $\tilde{A}\tilde{x} = \tilde{b}$:

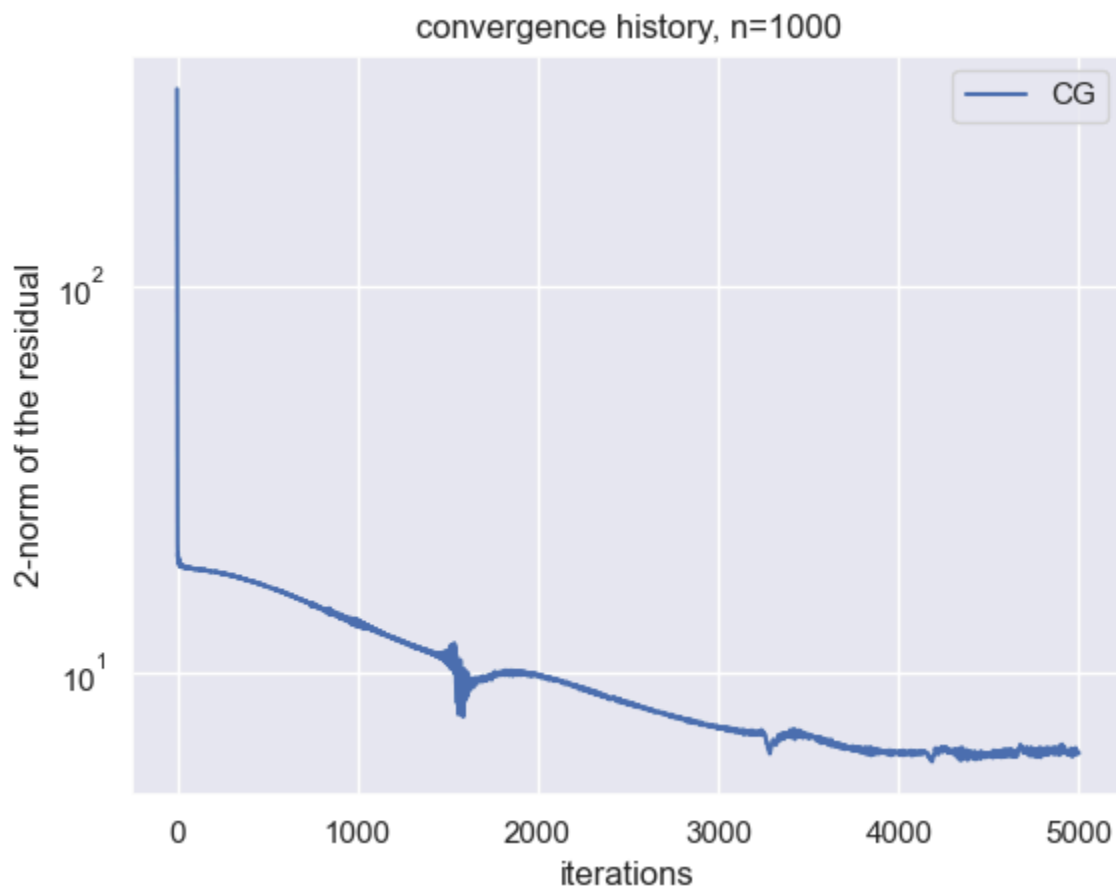
$$\begin{aligned}\tilde{r}_0 &= \tilde{b} - \tilde{A}\tilde{x}, \\ \tilde{p}_0 &= \tilde{r}_0, \\ \alpha_k &= \frac{\tilde{r}_k^T \tilde{r}_k}{\tilde{p}_k^T \tilde{A}\tilde{p}_k}, \\ \tilde{x}_{k+1} &= \tilde{x}_k + \alpha_k \tilde{p}_k, \\ \tilde{r}_{k+1} &= \tilde{r}_k - \alpha_k \tilde{A}\tilde{p}_k, \\ \beta_k &= \frac{\tilde{r}_{k+1}^T \tilde{r}_{k+1}}{\tilde{r}_k^T \tilde{r}_k}, \\ \tilde{p}_{k+1} &= \tilde{r}_{k+1} + \beta_k \tilde{p}_k.\end{aligned}$$

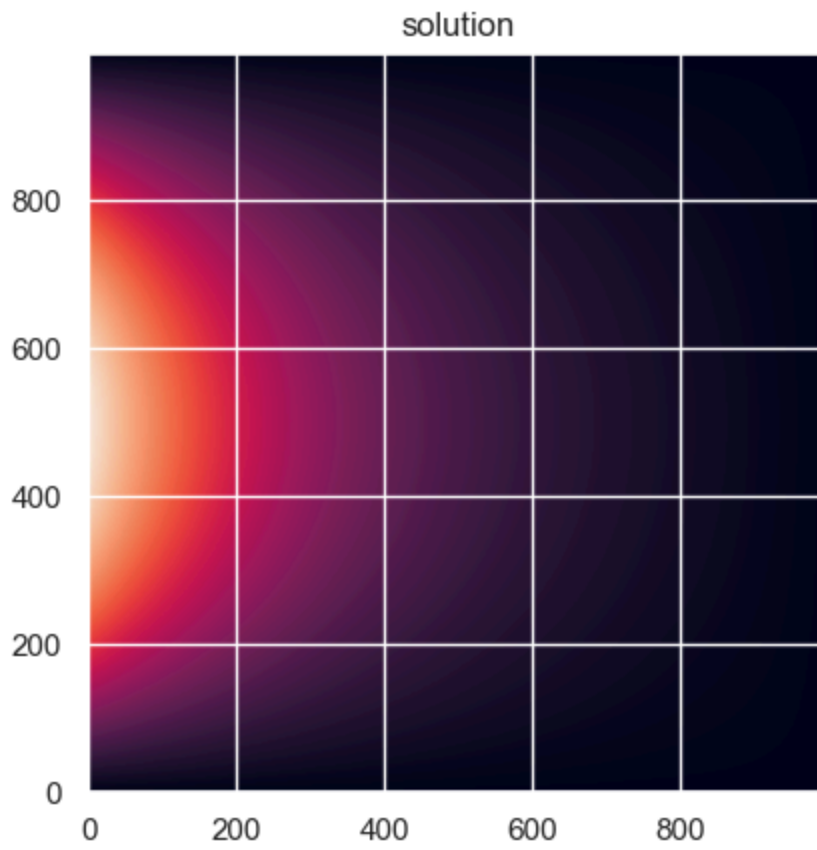
Substitute by $\tilde{x}_k = Cx_k, \tilde{r}_k = C^{-1}r_k, \tilde{p}_k = Cp_k$, we have

$$\begin{aligned}
r_0 &= b - Ax_0, \\
p_0 &= M^{-1}r_0, \\
\alpha_k &= \frac{r_k^T M^{-1}r_k}{p_k^T A p_k}, \\
x_{k+1} &= x_k + \alpha_k p_k, \\
r_{k+1} &= r_k - \alpha_k A p_k, \\
\beta_k &= \frac{r_{k+1}^T M^{-1}r_{k+1}}{r_k^T M^{-1}r_k}, \\
p_{k+1} &= M^{-1}r_{k+1} + \beta_k p_k.
\end{aligned}$$

Problem 3

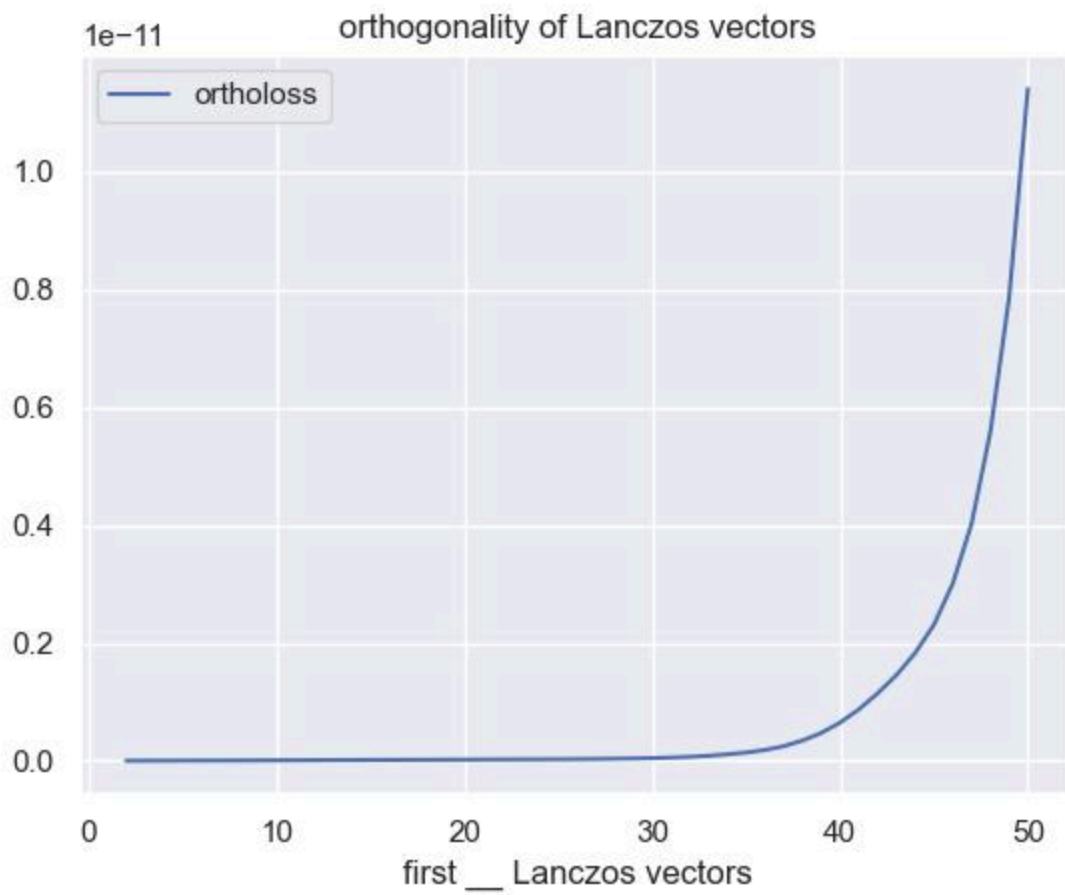
see `prob03.py` .

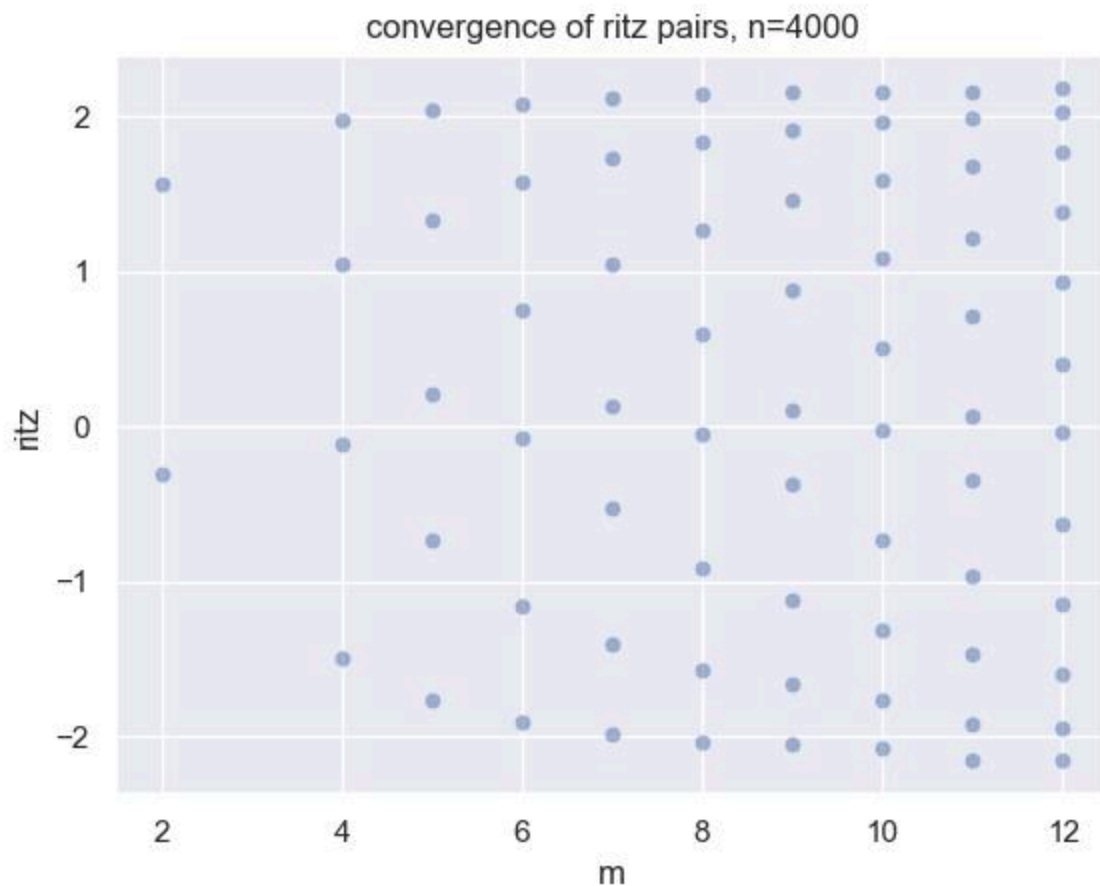




Problem 4

see `prob04.py` .





Problem 5

Since $f(A)$ is also symmetric,

$$(v - u)^T f(A)(v - u) = v^T f(A)v + u^T f(A)u - 2u^T f(A)v,$$

$$(v + u)^T f(A)(v + u) = v^T f(A)v + u^T f(A)u + 2u^T f(A)v,$$

then we can compute $u^T f(A)v$ by

$$u^T f(A)v = \frac{1}{4}((v + u)^T f(A)(v + u) - (v - u)^T f(A)(v - u)).$$