

However, if A is a symmetric (or Hermitian) matrix, then H_n is a symmetric (resp. Hermitian) tridiagonal matrix and more precise results can be shown; see Demmel [48] (Chapter 7, especially Section 7.2). We will consider the symmetric (and Hermitian) case in the next section, but first we show how Arnoldi iteration can be used to find approximations for the solution of a linear system $Ax = b$ where A is invertible but of very large dimension m .

18.5 GMRES

Suppose A is an invertible $m \times m$ matrix and let b be a nonzero vector in \mathbb{C}^m . Let $x_0 = A^{-1}b$, the unique solution of $Ax = b$. It is not hard to show that $x_0 \in \mathcal{K}_n(A, b)$ for some $n \leq m$. In fact, there is a unique monic polynomial $p(z)$ of minimal degree $s \leq m$ such that $p(A)b = 0$, so $x_0 \in \mathcal{K}_s(A, b)$. Thus it makes sense to search for a solution of $Ax = b$ in Krylov spaces of dimension $m \leq s$. The idea is to find an approximation $x_n \in \mathcal{K}_n(A, b)$ of x_0 such that $r_n = b - Ax_n$ is minimized, that is, $\|r_n\|_2 = \|b - Ax_n\|_2$ is minimized over $x_n \in \mathcal{K}_n(A, b)$. This minimization problem can be stated as

$$\text{minimize } \|r_n\|_2 = \|Ax_n - b\|_2, \quad x_n \in \mathcal{K}_n(A, b).$$

This is a least-squares problem, and we know how to solve it (see Section 23.1). The quantity r_n is known as the *residual* and the method which consists in minimizing $\|r_n\|_2$ is known as GMRES, for *generalized minimal residuals*.

Now since (u_1, \dots, u_n) is a basis of $\mathcal{K}_n(A, b)$ (since $n \leq s$, no breakdown occurs, except for $n = s$), we may write $x_n = U_n y$, so our minimization problem is

$$\text{minimize } \|AU_n y - b\|_2, \quad y \in \mathbb{C}^n.$$

Since by $(*)_1$ of Section 18.4, we have $AU_n = U_{n+1}\tilde{H}_n$, minimizing $\|AU_n y - b\|_2$ is equivalent to minimizing $\|U_{n+1}\tilde{H}_n y - b\|_2$ over \mathbb{C}^n . Since $U_{n+1}\tilde{H}_n y$ and b belong to the column space of U_{n+1} , minimizing $\|U_{n+1}\tilde{H}_n y - b\|_2$ is equivalent to minimizing $\|\tilde{H}_n y - U_{n+1}^* b\|_2$. However, by construction,

$$U_{n+1}^* b = \|b\|_2 e_1 \in \mathbb{C}^{n+1},$$

so our minimization problem can be stated as

$$\text{minimize } \|\tilde{H}_n y - \|b\|_2 e_1\|_2, \quad y \in \mathbb{C}^n.$$

The approximate solution of $Ax = b$ is then

$$x_n = U_n y.$$

Starting with $u_1 = b/\|b\|_2$ and with $n = 1$, the GMRES method runs $n \leq s$ Arnoldi iterations to find U_n and \tilde{H}_n , and then runs a method to solve the least squares problem

$$\text{minimize } \|\tilde{H}_n y - \|b\|_2 e_1\|_2, \quad y \in \mathbb{C}^n.$$