1. Write a function for reduction of a matrix $A$ to Hessenberg form, $H=Q^{*} A Q$. The function should both $H$ and $Q$. Test your function by applying it to a symmetric matrix (it should return a diagonal $H$ ), by checking that $Q$ is unitary, that $\left\|H-Q^{*} A Q\right\|$ is small, and by comparing the eigenvalues of $H$ to the eigenvalues of $A$ (they should be the same).
2. Write a function that computes an eigenvalue, eigenvector pair of a symmetric matrix $A$ via Rayleigh quotient iteration, given an initial guess for the eigenvector. Test the function with a random symmetric matrix $A$ reduced to Hessenberg form $H$ and a random initial guess for the eigenvector, and plotting $\left\|v^{(k+1)}-\left( \pm q_{J}\right)\right\|$ and $\left|\lambda^{(k)}-\lambda_{J}\right|$ versus $k$ on a log-linear plot (log errors on vertical, linear $k$ on horizontal), where $q_{J}$ and $\lambda_{J}$ are the eigenvector and eigenvalue of $H$ that the algorithm ultimately converges on (following Trefethen's notation in chapter 27).

Do the convergence rates match the cubic convergence predicted by Theorem 27.3?
3 (a) Implement the Arnoldi iteration algorithm to estimate the eigenvalues of a matrix $A$ iteratively. Test it against a random matrix $A$ with eigenvalues distributed over the unit circle, plotting the error versus iteration number $k$ of the first few Arnoldi estimates compared to the first few leading eigenvalues of $A$.
(b) Now set a few diagonal elements of $A$ to have magnitude $O(2)$ to $O(10)$ and recompute. What changes with the convergence rate? Why?
4. SVD problem still in production.

