## EE 636: Matrix Computations

## Syllabus

- Fundamentals: Flops count, memory management, matrix-vector multiplication.
- Gaussian elimination: Basic Gaussian elimination without pivoting, LU decomposition, the condition of LU decomposability. Gaussian elimination with pivoting. Positive definite matrices, Cholesky factorization. Comparison between the two methods. A brief discussion on sparsity.
- Sensitivity and round-off errors: Vector norms, matrix norms. Condition number. Perturbation, residual, round-off errors. Backward stability. Error propagation in Gaussian elimination. Backward error analysis in Gaussian elimination.
- **QR decomposition:** Orthogonal matrices, rotators and reflectors. Solution of the least squares problem, the full-rank case. Gram-Schmidt process. Condensed QR decomposition. Updating the QR decomposition.
- SVD: Introduction. Algorithm. Applications. Sensitivity.
- Eigenvalues and eigenvectors: The power method. Unitary similarity transform, Schur's theorem, normal matrices, spectral theorem of normal matrices. Hessenberg and tri-diagonal matrices, reduction to these forms. The QR algorithm. A brief discussion on sparsity.
- Iterative methods: Steepest descent, conjugate gradient.

## Textbooks

- David S. Watkins, *Fundamentals of Matrix Computions*, 3rd Edition, Willey-Interscience, 2010.
- Gene H. Golub and Charles F. Van Loan, *Matrix Computations*, 4th Edition, The Johns Hopkins University Press, 2013.

## Grading policy

- Two quizzes: 10% + 10%.
- Midsem: 20%.
- Endsem: 50%.
- Assignments and class performance: 10%.