

Homework 7, CS 6210, Fall 2020

Instructor: Austin R. Benson

Due December, 16, 2020 at 10:19am ET on CMS (before lecture)

## Policies

**Collaboration.** You are encouraged to discuss and collaborate on the homework, but you have to write up your own solutions and write your own code.

**Programming language.** You can use any programming language for the coding parts of the assignment. Code snippets that we provide and demos in class will use Julia.

**Typesetting.** Your write-up should be typeset with L<sup>A</sup>T<sub>E</sub>X. Handwritten homeworks are not accepted.

**Submission.** Submit your write-up as a single PDF on CMS: <https://cmsx.cs.cornell.edu>.

## Problems

### 1. *Augmented Lanczos.*

Consider the linear system

$$\begin{bmatrix} \lambda I & A \\ A^T & \lambda' I \end{bmatrix} \begin{bmatrix} s \\ x \end{bmatrix} = \begin{bmatrix} b \\ 0 \end{bmatrix},$$

where  $\lambda$  and  $\lambda'$  are scalars. Let  $T_{2k}$  and  $Q_{2k}$  be the matrices generated at step  $2k$  of Lanczos for this matrix.

(a) Show that

- the diagonal of  $T_{2k}$  is  $[\lambda \ \lambda' \ \lambda \ \lambda' \ \cdots \ \lambda \ \lambda']$ ,
- the super-diagonal of  $T_{2k}$  is  $[\beta_1 \ \beta'_1 \ \beta_2 \ \beta'_2 \ \cdots \ \beta_k]$ , and
- $Q_{2k} = \begin{bmatrix} u_1 & 0 & u_2 & 0 & \cdots & u_k & 0 \\ 0 & u'_1 & 0 & u'_2 & \cdots & 0 & u'_k \end{bmatrix}$ ,

where  $\beta_j, \beta'_j$  and  $u_j, u'_j$  are the same for any  $\lambda, \lambda'$ .

(b) Show that when  $\lambda' = -\lambda$ , the vector  $x$  solves the regularized least squares problem

$$\min_x \|Ax - b\|_2^2 + \lambda^2 \|x\|_2^2.$$

Choosing different values of  $\lambda$  and  $\lambda'$  and deriving CG or MINRES recurrences results in several iterative solvers. For example, LSQR corresponds to using CG for the special case in part (b).

### 2. *Searching for eigenpairs in Krylov subspaces.*

Show that  $K_k(A, b) = K_k(A - \sigma I, b)$  for any shift  $\sigma$ .

Thus, if we are looking for eigenvectors in a Krylov subspace, we can find approximations not only of eigenvectors corresponding to the largest-magnitude eigenvalues but to any extremal eigenpairs.

### 3. *Numerical experiments.*

Try out some of the built-in iterative solvers for finding the eigenpairs corresponding to the largest or smallest magnitude eigenvalues of a large sparse matrix.

Here is a Julia code snippet to help you start.

```

1 using Arpack
2 using SparseArrays
3
4 function eigensolver_timings()
5     A = sprand(5000, 5000, 0.01)
6     k = 10
7     # See interface at https://julialinearalgebra.github.io/Arpack.jl/latest/
8     t_LM = @elapsed eigs(A, nev=k, maxiter=5000, tol=1e-6, which=:LM)
9     t_SM = @elapsed eigs(A, nev=k, maxiter=5000, tol=1e-6, which=:SM)
10    return (t_LM, t_SM)
11 end

```

- (a) Does finding the largest or smallest magnitude eigenvalues take longer? Why might this be the case?
- (b) Make a plot that shows the running times of the solvers as a function of the number of eigenpairs  $k$ .

#### 4. Course evaluation (ungraded)

Please fill out the course evaluation. Also, feel free to directly email the course staff with feedback.

Thanks for a great semester, and good luck on the final! ☺