

# LECTURE NOTES FOR MTH 371, WINTER 2022

## "LARGE SCALE DATA ALGORITHMS"

PANAYOT S. VASSILEVSKI

### 1. HOMEWORK # 5: DUE MARCH 9

**Problem 1.1** (Preconditioned conjugate gradient (PCG) method).

- (i) *Implement the PCG algorithm (see below) which takes as input a  $n \times n$  s.p.d. matrix  $A$  (or a function that computes its actions), a preconditioner  $B$  that is either given as a s.p.d. matrix, or as a function that takes as input  $\mathbf{r}$  and returns  $\bar{\mathbf{r}} = B^{-1}\mathbf{r}$  which is the solution of  $B\bar{\mathbf{r}} = \mathbf{r}$ . Examples for  $B$  are the  $\ell_1$ -smoother, the symmetric Gauss-Seidel, and a two-level algorithm. The complete set of input parameters for the PCG algorithm are given in the algorithm below.*
- (ii) *Run the PCG algorithm for three preconditioners: (i)  $B$  being the diagonal of  $A$ , (ii)  $B$  being the symmetric Gauss-Seidel algorithm, (iii)  $B$  being defined by a symmetrized two-level algorithm (see below).*

**Algorithm 1.1** (Preconditioned Conjugate Gradient (PCG) Method).

*Given an s.p.d. matrix  $A$  (via its actions), r.h.s.  $\mathbf{b}$ , initial iterate  $\mathbf{x}$  (zero or a random vector), tolerance  $\epsilon$  ( $= 10^{-6}$ ) and a maximal number of iterations,  $\text{max\_iter}$  ( $= 1000$ ). For a given s.p.d. preconditioner  $B$ , which computes  $B^{-1}\mathbf{r}$  for any given vector  $\mathbf{r}$ , we perform:*

- *Initiate: compute initial residual  $\mathbf{r} = \mathbf{b} - A\mathbf{x}$ , the preconditioned residual  $\bar{\mathbf{r}} = B^{-1}\mathbf{r}$ , the squared norm  $\delta_0 = \mathbf{r}^T B^{-1}\mathbf{r} = \|\mathbf{r}\|_{B^{-1}}^2 = \mathbf{r}^T \bar{\mathbf{r}}$ ,  $\delta = \delta_0$ , and let  $\mathbf{p} = \bar{\mathbf{r}}$  be the initial search vector.*
- *Loop: for  $\text{iter} = 1, 2, \dots, \text{max\_iter}$  compute*
  - $\mathbf{g} = A\mathbf{p}$ .
  - $\alpha = \frac{\mathbf{r}^T B^{-1}\mathbf{r}}{\mathbf{p}^T A\mathbf{p}} = \frac{\delta}{\mathbf{p}^T \mathbf{g}}$ .
  - *Update iterate  $\mathbf{x} := \mathbf{x} + \alpha\mathbf{p}$ .*
  - *Update residual  $\mathbf{r} = \mathbf{r} - \alpha A\mathbf{p} = \mathbf{r} - \alpha\mathbf{g}$ .*
  - *New preconditioned residual:  $\bar{\mathbf{r}} = B^{-1}\mathbf{r}$ .*
  - *Let  $\delta_{\text{old}} = \delta$ .*
  - *Compute  $\delta = \mathbf{r}^T B^{-1}\mathbf{r} = \mathbf{r}^T \bar{\mathbf{r}}$ .*
  - *Check for convergence: If  $(\delta < \epsilon^2 \delta_0)$  exit due to achieved convergence tolerance.*
  - *Compute next search direction:*
    - \* *Compute  $\beta = \frac{\mathbf{r}^T B^{-1}\mathbf{r}}{\mathbf{r}_{\text{old}}^T B^{-1}\mathbf{r}_{\text{old}}} = \frac{\delta}{\delta_{\text{old}}}$ .*
    - \* *Next search direction is:  $\mathbf{p} = \bar{\mathbf{r}} + \beta\mathbf{p}$ .*
- *End of Loop on iter.*

**Algorithm 1.2** (Symmetric two-level algorithm).

Let  $A$  be a  $n \times n$  s.p.d. matrix. Given a convergent method based on a matrix  $M$  (either  $\ell_1$ -smoother or the forward Gauss-Seidel) and consider  $M^T$  as well (it will be either the  $\ell_1$ -smoother, i.e.,  $M^T = M$ , or the backward Gauss-Seidel, if  $M$  is the forward Gauss-Seidel). We also need the matrix  $P$  corresponding to a set of aggregates, i.e., the relation "vertex\_aggregate" constructed for the sparsity graph of the sparse matrix  $A$  for example by the Luby's algorithm. We also need the coarse matrix  $A_c = P^T A P$ . The two-level algorithm requires function that solves equations with  $A_c$  (use the  $LDL^T$  or LU factorization of the s.p.d. matrix  $A_c$ ).

The symmetric two-level algorithm takes as input a vector  $\mathbf{r}$  and provides on output  $\bar{\mathbf{r}} = B^{-1}\mathbf{r}$ , which is defined by the following steps;

- Solve for  $\mathbf{y}$ ,  $M\mathbf{y} = \mathbf{r}$ .
- Compute coarse r.h.s.  $\mathbf{r}_c = P^T(\mathbf{r} - A\mathbf{y})$ .
- Solve the coarse problem for  $\mathbf{y}_c$ ,  $A_c\mathbf{y}_c = \mathbf{r}_c$  (use  $LDL^T$  or LU facorization of  $A_c$ ).
- Update  $\mathbf{y} = \mathbf{y} + P\mathbf{y}_c$ .
- Solve for correction  $\mathbf{z}$ ,  $M^T\mathbf{z} = \mathbf{r} - A\mathbf{y}$ .
- Update  $\mathbf{y} := \mathbf{y} + \mathbf{z}$ .
- $\bar{\mathbf{r}} = \mathbf{y}$ .

The above mapping  $B^{-1} : \mathbf{r} \mapsto \bar{\mathbf{r}}$  is referred to as the two-level preconditioner, and  $B$  is sometimes denoted  $B_{TL}$ .

FARIBORZ MASEEH DEPARTMENT OF MATHEMATICS AND STATISTICS, PORTLAND STATE UNIVERSITY, PORTLAND, OR 97201

Email address: panayot@pdx.edu