## Homework 4 Foundations of Computational Math 2 Spring 2021

These study questions relate to the material optimization-based iterative methods for linear systems defined by symmetric positive definite matrices.

## Problem 4.1

Consider the minimization problem

$$
\min _{x \in \mathbb{R}^{n}} f(x)
$$

where $f(x)=\frac{1}{2} x^{T} A x-x^{T} b, A \in \mathbb{R}^{n \times n}$ is symmetric positive definite, and $b \in \mathbb{R}^{n}$.
(4.1.a) Show that $\forall 0 \leq \beta \leq 1$

$$
\beta f(x) \geq f(\beta x)
$$

(4.1.b) Suppose $x_{0} \in \mathbb{R}^{n}, x_{1} \in \mathbb{R}^{n}$ and $x_{0} \neq x_{1}$. Show that $f(x)$ is a convex function.

## Problem 4.2

Let $A=Q \Lambda Q^{T}$ be a symmetric positive definite matrix where $Q$ is an orthogonal matrix and $\Lambda$ is a diagonal matrix whose diagonal elements are positive and also are the eigenvalues of $A$. Define

$$
\begin{array}{rll}
\tilde{x}=Q^{T} x & \text { and } & \tilde{b}=Q^{T} b \\
A x=b & \text { and } & \Lambda \tilde{x}=\tilde{b}
\end{array}
$$

Given $x_{0}$ and $\tilde{x}_{0}$, define the sequence $x_{k}$ as the sequence of vectors produced by steepest descent applied to $A x=b$ and the sequence $\tilde{x}_{k}$ as the sequence of vectors produced by steepest descent applied to $\Lambda \tilde{x}=\tilde{b}$.

Let $e_{k}=x_{k}-x$ and $\tilde{e}_{k}=\tilde{x}_{k}-\tilde{x}$. Show that if $\tilde{x}_{0}=Q^{T} x_{0}$ then

$$
\left\|e_{k}\right\|_{2}=\left\|\tilde{e}_{k}\right\|_{2}, \quad k>0
$$

## Problem 4.3

Let $A \in \mathbb{R}^{n \times n}$ be symmetric postive definite with an eigendecompositon $A=Q \Lambda Q^{T}$ with $Q \in \mathbb{R}^{n \times n}$ and orthogonal matrix, i.e., $Q^{T} Q=Q Q^{T}=I$, and $\Lambda \in \mathbb{R}^{n \times n}$ a diagonal matrix with positive diagonal elements $\lambda_{i}=e_{i}^{T} \Lambda e_{i}>0$.

Consider the two systems $A x=b$ and $\Lambda \tilde{x}=\tilde{b}$ with $Q \tilde{x}=x$ and $Q \tilde{b}=b$. The iterations defined by applying Steepest Descent (SD) to each are

$$
\begin{aligned}
& x_{k+1}=x_{k}+\alpha_{k} r_{k}, \quad r_{k}=b-A x_{k}, \quad \alpha_{k}=\frac{r_{k}^{T} r_{k}}{r_{k}^{T} A r_{k}} \\
& \tilde{x}_{k+1}=\tilde{x}_{k}+\tilde{\alpha}_{k} \tilde{r}_{k}, \quad \tilde{r}_{k}=\tilde{b}-\Lambda \tilde{x}_{k}, \quad \tilde{\alpha}_{k}=\frac{\tilde{r}_{k}^{T} \tilde{r}_{k}}{\tilde{r}_{k}^{T} \Lambda \tilde{r}_{k}}
\end{aligned}
$$

given $x_{0}$ and $Q \tilde{x}_{0}=x_{0}$. The elements of the vectors with the tildes are the coefficients of the corresponding vectors without the tildes with respect to the basis of eigenvectors given by the columns of $Q$.
(4.3.a) Show that $\alpha_{k}=\tilde{\alpha}_{k}$ and that

$$
\alpha_{k}^{-1}=\tilde{\alpha}_{k}^{-1}=\sum_{i=1}^{n} \gamma_{i} \lambda_{i}, \quad \gamma_{i} \geq 0, \quad \sum_{i=1}^{n} \gamma_{i}=1
$$

(4.3.b) Any $x_{0} \in \mathbb{R}^{n}$ can be corrected to $A^{-1} b$ by

$$
A^{-1} b=x_{0}+c_{0}, \quad c_{0}=A^{-1}\left(b-A x_{0}\right)=A^{-1} r_{0} .
$$

Consider applying SD to $A x=b$. Derive a sufficient condition on $A$ so that for any $x_{0}$ convergence to $A^{-1} b$ occurs in one step, i.e.,

$$
A^{-1} b=x_{1}=x_{0}+\alpha_{0} r_{0}
$$

(4.3.c) Is the condition also a necessary condition for convergence of SD in one step for any $x_{0}$ ?

## Problem 4.4

## 4.4.a

Consider the iteration:

$$
\begin{aligned}
y_{0} & =0 \\
y_{i+1} & =y_{i}+\tilde{\alpha}_{i} e_{i+1} \\
\tilde{r}_{i} & =b-D y_{i} \\
\tilde{\alpha}_{i} & =\frac{e_{i+1}^{T} \tilde{r}_{i}}{e_{i+1}^{T} D e_{i+1}}
\end{aligned}
$$

where $D \in \mathbb{R}^{n \times n}$ is a nonsingular diagonal matrix.
Show that $y_{n}=y=D^{-1} b$.

## 4.4.b

Suppose $A \in \mathbb{R}^{n \times n}$ is a symmetric positive definite matrix and $A x=b$. Let $p_{1}, p_{2}, \cdots, p_{n}$ be real vectors that are $A$-orthogonal, i.e., $<p_{i}, p_{j}>=0$ if $i \neq j$ and $<p_{i}, p_{i} \gg 0$ where $\langle w, v\rangle=w^{T} A v$ is the inner product on $\mathbb{R}^{n}$ defined by $A$.

Use the result from the first part of the problem to show that the conjugate direction iteration:

$$
\begin{aligned}
x_{0} & =0 \\
x_{i+1} & =x_{i}+\alpha_{i} p_{i+1} \\
r_{i} & =b-A x_{i} \\
\alpha_{i} & =\frac{p_{i+1}^{T} r_{i}}{p_{i+1}^{T} A p_{i+1}}
\end{aligned}
$$

is such that $x_{n}=x=A^{-1} b$.

## Problem 4.5

The conjugate direction iteration (CD) can also be derived from a basis expansion point of view. Let $e_{\text {true }}=x-x_{0}=A^{-1} b-x_{0}$ where $A$ is a symmetric positive definite matrix. Let $<w, v\rangle=w^{T} A v$ be the inner product on $\mathbb{R}^{n}$ defined by $A$ and $p_{1}, p_{2}, \cdots, p_{n}$ be real vectors that are $A$-orthonormal, i.e., $<p_{i}, p_{j}>=0$ if $i \neq j$ and $<p_{i}, p_{i}>=1$.
4.5.a Show that any vector can be easily written in terms of a basis that is orthonormal with respect to some inner product and apply this to $e_{\text {true }}$ to get

$$
\begin{equation*}
e_{\text {true }}=p_{1}<p_{1}, e_{\text {true }}>+\cdots p_{n}<p_{n}, e_{\text {true }}> \tag{1}
\end{equation*}
$$

4.5.b Show that for any $x_{0}$

$$
\begin{aligned}
\alpha_{i} & =p_{i+1}^{T} r_{i} \\
x_{i+1} & =x_{i}+\alpha_{i} p_{i+1} \\
r_{i} & =b-A x_{i}
\end{aligned}
$$

is such that $x_{n}=x=A^{-1} b$.
Hint: Define an iteration based on (1) that yields $x_{n}=x$ and then show it can be computed via the CD iteration given in this problem.

## Problem 4.6

Recall the basic CD/CG properties that hold given the assumption that CG has not converged at step $k$,

- $x_{k}=\alpha_{0} d_{0}+\cdots+\alpha_{k-1} d_{k-1}$ is optimal (inherited from CD), i.e.,

$$
\forall x \in x_{0}+\operatorname{span}\left[d_{0}, d_{1}, \ldots, d_{k-1}\right], \quad\left\|x_{k}-A^{-1} b\right\|_{A} \leq\left\|x-A^{-1} b\right\|_{A}
$$

- $<d_{k}, d_{j}>_{A}=0 i \neq j$ for $0 \leq i, j \leq k-1$ (inherited from CD).
- $<r_{k}, d_{j}>=0$ for $0 \leq j \leq k-1$ (inherited from CD).
- $\left\langle r_{k}, r_{j}>=0\right.$ for $0 \leq j \leq k-1$ (CG-specific).
- $\operatorname{span}\left[d_{0}, d_{1}, \ldots, d_{k}\right]=\operatorname{span}\left[r_{0}, r_{1}, \ldots, r_{k}\right]$ (CG-specific).
- $\operatorname{span}\left[r_{0}, r_{1}, \ldots, r_{k}\right]=\operatorname{span}\left[r_{0}, A r_{0}, \ldots, A^{k} r_{0}\right]$ (CG-specific).

Given the inherited properties prove the three CG-specific properties.

## Problem 4.7

## 4.7.a

Let the cost function $f: \mathbb{R}^{2} \rightarrow \mathbb{R}$ be defined by

$$
f(x)=x^{T} d+x^{T} x, \quad \text { where } \quad d=\binom{\delta_{1}}{\delta_{2}}
$$

where $\delta_{1}>0, \delta_{2}>0$ and $\mu=\|d\|_{2}>1$. Consider the problem

$$
\min _{x \in \mathbb{R}^{2}} f(x) .
$$

(i) Find a minimizer $x^{*}$. Is it unique?
(ii) Write the iteration that defines applying the steepest descent algorithm to solve the minimization problem.
(iii) How would you set the stepsize $\alpha_{k}$ and why?
(iv) Will your choice of $\alpha_{k}$ yield an algorithm that converges in a finite number of steps?

## 4.7.b

Now suppose the minimization problem is constrained so that we are only interested in $x \in \mathbb{R}^{2}$ on the circle of radius 1 , i.e., the unit circle

$$
\mathcal{S}_{1}=\left\{x \in \mathbb{R}^{2} \mid x^{T} x=1\right\}
$$

Specifically, we want to solve

$$
\min _{x \in \mathcal{S}_{1}} f(x)
$$

(i) Show that this problem can be viewed as an unconstrained minimization problem on $\mathbb{R}$ by writing the cost function over $\mathcal{S}_{1}$ as a function of a real variable $\theta$.
(ii) Write the iteration that defines applying the steepest descent algorithm to solve the minimization problem over $\mathbb{R}$.
(iii) How would you set the stepsize $\alpha_{k}$ and why?
(iv) Will your choice of $\alpha_{k}$ yield an algorithm that converges in a finite number of steps when started at an initial guess $\theta_{0}=0$ ?

