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# ON THE PENALTY APPROXIMATION OF QUADRATIC PROGRAMMING PROBLEM 

ZDENĚK DOSTÁL

An upper bound for the difference of the exact solution of the problem of minimization of quadratic functional on a subspace and its penalty approximation has been given. The paper is supplied with a numerical example.

## 1. INTRODUCTION

It is an easy consequence of the general theory of the penalty method [1] that the solution $\bar{x}$ of the problem

$$
\begin{equation*}
\min \{f(x): x \in \mathscr{V}\} \tag{1}
\end{equation*}
$$

where $f(x)=\frac{1}{2} x^{\mathrm{T}} A x-b^{\mathrm{T}} x$ is a convex quadratic functional and $\mathscr{V}$ is a subspace of the real Euclidean space $\mathbb{R}^{n}$, may be approximated by the solution $\bar{x}_{\varepsilon}$ of the problem

$$
\begin{equation*}
\min \left\{f(x)+\frac{1}{2} \varepsilon^{-1} x^{\mathrm{T}} P x: x \in \mathbb{R}^{n}\right\} \tag{2}
\end{equation*}
$$

where $\varepsilon>0$ and $P$ is any $n \times n$ matrix with the range $\mathscr{V}$.
Even though there is a number of exact methods for transformation of the problem (1) to the unconstrained one [3], hardly any of them is as simple and cheap from the point of view of numerical realisation as the penalty approximation (2). In a special case when $P$ is diagonal its application consists simply in overwriting corresponding diagonal entries of $A$ by a large number and is widely used [2]. In a more general case, when values of constrained variables are not fully determined by constraints, the application is complicated by the fact that adding a too large matrix to $A$ may destroy the corresponding part of $A$. This has motivated us to examine in detail the difference $\bar{x}-\bar{x}_{\varepsilon}$ and to carry out some numerical experiments in order to assess the applicability of the penalty approach.

## 2. AN ESTIMATE

Let $f(x)=\frac{1}{2} x^{\mathrm{T}} A x-b^{\mathrm{T}} x$ where $A \in \mathbb{R}^{n \times n}$ is a real positive semidefinite matrix and $b \in \mathbb{R}^{n}$. Denote by $\mathscr{V}$ and $\mathscr{U}$ a subspace of $\mathbb{R}^{n}$ and its orthogonal complement.

Let $V \in \mathbb{R}^{n \times m}$ and $U \in \mathbb{R}^{n \times(n-m)}$ be two matrices whose columns are formed by the orthonormal bases of $\mathscr{V}$ and $\mathscr{U}$, respectively. Denote by $P$ a conjugate projector on $\mathscr{V}$. Let us remind that a matrix $P$ is a conjugate projector on a subspace $\mathscr{V}$ iff $P^{2}=P$ and $P^{\mathrm{T}} A(I-P)=0$. In this case $I-P$ is a conjugate projector on the null space of $P$. Notice that $A P$ is symmetric and positive semidefinite as $P^{\mathrm{T}} A=P^{\mathrm{T}} A P$.

Theorem. Let $f, U, V . \mathscr{V}$ be defined as above, let $\bar{x}$ be the solution of the problem $\min _{\gamma} f(x)$, and let $\bar{x}_{\varepsilon}$ denote for each $\varepsilon>0$ the solution of the problem $\min (f(x)+$ $\left.+\frac{1}{2} \varepsilon^{-1} x^{\mathrm{T}} U U^{\mathrm{T}} x\right)$.

If $V^{\mathrm{T}} A V$ is invertible, then

$$
\begin{equation*}
\left|\bar{x}-\bar{x}_{\varepsilon}\right| \leqq \varepsilon\left(\left|P-V V^{\mathrm{T}}\right|+1\right)^{2}|b|, \tag{3}
\end{equation*}
$$

where $P=V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A$ is a conjugate projector on $\mathscr{V}$.
Proof. Under the assumption of the theorem, obviously $\bar{x}=V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{r}} b$ and $A+\varepsilon^{-1} U U^{\mathrm{T}}$ is positive definite.

Now notice that $Q=(V, U)$ is an $n \times n$ orthogonal matrix, so that

$$
\left(A+\varepsilon^{-1} U U^{\mathrm{T}}\right)^{-1}=Q\left(Q^{\mathrm{T}}\left(A+\varepsilon^{-1} U U^{\mathrm{T}}\right) Q\right)^{-1} Q^{\mathrm{T}}
$$

As

$$
Q^{\mathrm{T}}\left(A+\varepsilon^{-1} U U^{\mathrm{T}}\right) Q=\left(\begin{array}{ll}
V^{\mathrm{T}} A V, & V^{\mathrm{T}} A U \\
U^{\mathrm{T}} A V, & U^{\mathrm{T}} A U+\varepsilon^{-1} I
\end{array}\right)
$$

we can apply the formula for the inverse of $2 \times 2$ block matrix to get

$$
\begin{align*}
& \bar{x}_{\varepsilon}=(V, U)\left(\begin{array}{cc}
\left(V^{\mathrm{T}} A V\right)^{-1}+C Z C^{\mathrm{T}}, & -C Z \\
-Z C^{\mathrm{T}} & Z
\end{array}\right)\binom{V^{\mathrm{T}}}{U^{\mathrm{T}}} b= \\
& =\bar{x}+V C Z^{\mathrm{T}} V^{\mathrm{T}} b-V C Z U^{\mathrm{T}} b-U Z C^{\mathrm{T}} V^{\mathrm{T}} b+U Z U^{\mathrm{T}} b, \tag{4}
\end{align*}
$$

where

$$
C=\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A U
$$

and

$$
Z=\left(U^{\mathrm{T}} A U+\varepsilon^{-1} I-U^{\mathrm{T}} A V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A U\right)^{-1}
$$

Let us now estimate the norm of $Z$. First notice that

$$
\begin{equation*}
Z=\varepsilon\left(I-\varepsilon U^{\mathrm{T}} A\left(V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A-I\right) U\right)^{-1} \tag{5}
\end{equation*}
$$

Since $P=V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A$ is a conjugate projector on $\mathscr{V}$, we can rewrite $Z$ into the form

$$
Z=\varepsilon\left(I+\varepsilon U^{\mathrm{T}} A(I-P) U\right)^{-1}
$$

where the inverted matrix is obviously positive definite with all the eigenvalues greater or equal to 1 . It follows that

$$
|Z| \leqq \varepsilon
$$

Further, notice that

$$
\begin{aligned}
& |V C|=\left|V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A U\right|=|P U|=\left|P\left(I-V V^{\mathrm{T}}\right)\right|= \\
& =\left|P-P V V^{\mathrm{T}}\right|=\left|P-V V^{\mathrm{T}}\right|
\end{aligned}
$$

Taking into account that both $|U| \leqq 1$ and $|V| \leqq 1$, we get

$$
\begin{aligned}
& \left|x-x_{\varepsilon}\right|=\left|V C Z C^{\mathrm{T}} V^{\mathrm{T}} b-V C Z U^{\mathrm{T}} b-U Z C^{\mathrm{T}} V^{\mathrm{r}} b+U Z U^{\mathrm{T}} b\right| \leqq \\
& \leqq\left(|V C|^{2}+2|V C|+1\right)|Z||b| \leqq \varepsilon(|V C|+1)^{2}|b|= \\
& =\varepsilon\left(\left|P-V V^{\mathrm{T}}\right|+1\right)^{2}|b|
\end{aligned}
$$

Consider the following trivial example: $A=I, b=(1,0 \ldots)^{\mathrm{T}}, \mathscr{V}$ is the set of all vectors with the first coordinate equal to 0 . Then obviously $|b|=1$,

$$
\left|\bar{x}-\bar{x}_{\varepsilon}\right|=\varepsilon(1+\varepsilon)^{-1},
$$

while (3) yields

$$
\left|\bar{x}-\bar{x}_{\varepsilon}\right| \leqq \varepsilon .
$$

This shows that the constant $\left(\left|P-V V^{\mathrm{T}}\right|+1\right)^{2}$ can not be improved without additional assumptions.

The following corollary gives us an idea about the constant in (3) in terms of the spectrum of $A$.

Corollary. Let $f, U, V, A, \bar{x}$ and $\bar{x}_{\varepsilon}$ be those of the theorem, and let $A$ be invertible. Then

$$
\begin{equation*}
\left|\bar{x}-\bar{x}_{\varepsilon}\right| \leqq \varepsilon(\varkappa(A)+1)^{2}|b| \tag{4}
\end{equation*}
$$

where $\chi(A)$ is the spectral condition number of $A$.
Proof. Notice that

$$
\left|P-V V^{\mathrm{T}}\right|=\left|\left(P-V V^{\mathrm{T}}\right)\left(U U^{\mathrm{T}}+V V^{\mathrm{T}}\right)\right|=\left|P U U^{\mathrm{T}}\right| \leqq|P|
$$

and

$$
|P|=\left|V\left(V^{\mathrm{T}} A V\right)^{-1} V^{\mathrm{T}} A\right| \leqq\left|\left(V^{\mathrm{T}} A V\right)^{-1}\right||A| \leqq\left|A^{-1}\right||A|
$$

Thus $\left|P-V V^{\mathrm{T}}\right| \leqq x(A)$, which substituted into (3) yields (4).

## 3. NUMERICAL EXAMPLE

As an illustration, consider the problem

$$
\min _{x \in \mathscr{r}} \frac{1}{2} x^{\mathrm{T}} A x-b^{\mathrm{T}} x
$$

where

$$
\begin{aligned}
& A=\left(\begin{array}{rrr}
2 & -1 & 0 \\
-1 & 2 & -1 \\
0 & -1 & 2
\end{array}\right), \quad b=\left(\begin{array}{l}
1 \\
1 \\
1
\end{array}\right), \quad x=\left(\begin{array}{l}
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right), \\
& \mathscr{V}=\left\{x: x_{1}+x_{3}=0\right\} .
\end{aligned}
$$

Direct computations show that $x(A)=(1+\sqrt{ } 2) /(1-\sqrt{ } 2)$, so that we get the estimate

$$
\left|x-x_{\varepsilon}\right| \leqq 80 \cdot 77 \varepsilon
$$

The values of both sides are tabulated in Table 1. All the penalty computations were carried out in double precision and no serious effect of computer arithmetics has

## Table 1.

| $\varepsilon$ | $10^{-1}$ | $10^{-4}$ | $10^{-7}$ |
| :--- | :---: | :---: | :---: |
| $\left\|x-x_{\varepsilon}\right\|$ | $1 \cdot 716 \mathrm{E}-1$ | $1.837 \mathrm{E}-4$ | $1.837 \mathrm{E}-7$ |
| $\varepsilon(1+x(A))^{2}\|b\|$ | 8.077 E 0 | $8.077 \mathrm{E}-3$ | $8.077 \mathrm{E}-6$ |

been observed with $\varepsilon^{-1} \leqq 10^{10}$. We conclude that the estimate (4) may give us some idea about the error of penalty computations, though in many cases it will be too pesimistic.
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