

UNIVERSITI PUTRA MALAYSIA

SCALED THREE-TERM CONJUGATE GRADIENT METHOD VIA DAVIDON-FLETCHER-POWELL UPDATE FOR UNCONSTRAINED OPTIMIZATION

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ARZUKA IBRAHIM

Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia, in Fulfilment of the Requirements for the Degree of Master of Science

May, 2015

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DEDICATIONS

To my Dad Late Alhaji Arzuka Jega



Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfilment of the requirement for the degree of Master of Science

SCALED THREE-TERM CONJUGATE GRADIENT METHOD VIA DAVIDON-FLETCHER-POWELL UPDATE FOR UNCONSTRAINED OPTIMIZATION

By

ARZUKA IBRAHIM

May 2015

Chair: Associate Professor Mohd Rizam Abu Bakar, PhD Faculty: Institute for Mathematical Research

This thesis focus on the development of Scaled Three-Term Conjugate Gradient Method via the Davidon-Fletcher-Powell (DFF) quasi-Newton update for unconstrained optimization. The DFP method possess the merits of Newton's method and steepest descent method while overcoming their disadvantages. Over the years the DFP update has been neglected as a result of lacking the self correcting property for bad Hessian approximation. In this thesis, we proposed a Scaled Three-Term Conjugate Gradient Method by utilizing the DFP update for the inverse Hessian approximation via memoryless quasi Newton's method which satisfies both the sufficient descent and the conjugacy conditions. The basic philosophy is to restart the DFP update with a multiple of identity matrix in every iteration. An acceleration scheme is incorporated in the proposed method to enhance reduction in function value. Numerical results from an implementation of the proposed method on some standard unconstrained optimization problem shows that the proposed method is promising and exhibits superior numerical performance in comparison with other well-known conjugate gradient methods.

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Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Master Sains

KAEDAH KECERUNAN KONJUGAT BERSKALA TIGA SEBUTAN MELALUI KEMASKINI DAVIDON-FLETCHER-POWELL UNTUK PENGOPTIMUNAN TAK BERKEKANGAN

Oleh

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Pengerusi: Profesor Madya Mohd Rizam Abu Bakar, PhD Fakulti: Institut Penyelidikan Matematik

Tumpuan tesis ini adalah terhadap pembentukan kaedah kecerunan konjugat berskala tiga sebutan melalui kemaskini kuasi-Newton Davidon-Fletcher-Powell (DFP) untuk pengoptimuman tak berkekangan. Kaedah Davidon-Fletcher-Powell memiliki merit kaedah Newton dan kaedah penurunan tercuram di samping berupaya mengatasi kelemahan kaedah-kaedah ini. Sejak beberapa tahun kebelakangan ini kemaskini DFP diabaikan kerana kelemahan sifat pembetulan sendiri untuk penghamiran Hessian yang lemah. Dalam tesis ini, kami mencadangkan kaedah kecerunan konjugat berskala tiga sebutan dengan mengounakan kemaskini DFP untuk penghampiran Hessian songsang melalui kaedah kuasi-Newton tanpa memori yang memenuhi syarat penurunan kecerunan yang mencukupi dan syarat konjugasi. Falsafah asas adalah untuk mula semula kemaskini DFP dengan beberapa matrik identiti dalam setiap lelaran. Skim pecutan digabungkan ke dalam kaedah yang dicadangkan untuk meng galakkan pengurangan nilai fungsi. Keputusan berangka dari pelaksanaan kaedah yang dicadang ke atas beberapa masalah pengoptimuman tak berkekangan piawai, menunjukkan bahawa kaedah ini berpotensi dan mempamerkan prestasi berangka unggul berbanding dengan kaedah-kaedah kecerunan konjugat lain yang terkenal.

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LIST OF ABBREVIATIONS

f	Objective function
α, α_k	Steplength (at k th iteration)
\Re^n	n-dimensional space
x	Vector
x_{k+1}, x_k	Current and previous point
$\nabla f(x_k)$	Gradient of the f at the point x_k
$\nabla^2 f(x_k)$	Second derivative of f at the point x_k Hessian
x^*	Minimizer of
$(\ .\ _2) x$	Is the norm of a vector or matrix
x	Variable of an optimization problem
Ι	Identity matrix
tr	Trace operator
NI	Number of iterations
NF	Number of function evaluation
SD	Steepest decent method
DFP	Davidon-Fletcher-Powell
SR1	Symmetric Rank One
BFGS	Broyden-Fletcher-Goldfarb-Shanno
STCG	Scaled three-term Conjugate Method
TTHS	Three-term Hestenes-Stiefel
TTPRP	Three-term Polaik-Reeve-Polaik
TTDFP	Three-term Davidon-Fletcher-Powell
MTTPRP	Modified Three-term Polaik-Reeve-Polaik

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CHAPTER 1

INTRODUCTION

1.1 Introduction

Numerous problems in the real-life situation possess several solutions, Ancona-Navarrete and Tawn (2000) in some cases infinite number of solution may be possible Antoniou and Lu (2007). Optimization is the process of locating the best out of the available solutions, it occurs in various fields of application such as engineering, sciences, social sciences and administration. It trace back to early century and become more independent in the late 1940's when G.B Dantzing suggested the famous simplex method. With the advent of digital computers, this area received great attention. Optimization methods are employed in order to obtain the most or the least in a given situation, these methods are iterative rules design to determine the optimal solution of a given optimization problem. In this thesis, we focus on solving nonlinear large-scale unconstrained optimization problem of the form

$$\min f(x), \ x \in \Re^n, \tag{1.1}$$

where $f: \Re^n \to \Re$ is a continuously differentiable function and n is the dimension of x_k which is assume to be large. Several iterative schemes were proposed to generate a sequence of approximation to the solution of (1.1). The basic idea behind optimization method is to either minimize or maximize the problem depending upon it nature. Given an initial point $x_0 \in \Re^n$, generate a sequences $\{x_k\}$ of approximation using some iterative scheme such that the iterate x_k moves toward the neighborhood of the optimal value and readily converges when a given stopping criterion is satisfied. In most situation $||g_k|| < \epsilon$ where ϵ is pre-determined tolerance, that is when the gradient approaches zero, the approximations x_k of the iterate converges to the optimal solution of the given problem. The sequences of approximations to the solution of (1.1) are generated by the following iterative scheme

$$x_{k+1} = x_k + \alpha_k d_k, \qquad k = 0, 1, 2, 3, \dots,$$
(1.2)

where x_{k+1} and x_k are the current and previous iterate, $\alpha_k > 0$ is the steplength determine by a line search strategy and d_k is the search direction. Different d_k will result to a different iterative scheme, the search direction are mainly classified into two classes. In the first class we have

$$d_k = -H_k g_k. \tag{1.3}$$

If (H_k) in (1.3) is an identity matrix then (1.3) corresponds to steepest descent direction and the iterative scheme (1.2) become steepest descent method which is one of the earliest method for solving (1.1). If $H_k = \nabla^2 f(x_k)$ in (1.3), then (1.3) corresponds to Newton's direction and the iterative scheme (1.2) becomes the Newton's method. Moreover if (H_k) in (1.3) is an approximation of $\nabla^2 f(x_k)$ then (1.3) corresponds to quasi-Newton direction and the iterative scheme (1.2) becomes quasi-Newton methods which were developed based on the shortcoming of the Newton method, they require the gradient of the objective function to generate and update the inverse Hessian approximate matrix at every step such that the secant condition

$$H_k y_k = s_k, \tag{1.4}$$

is satisfied, with $y_k = g_{k+1} - g_k$ and $s_k = x_{k+1} - x_k$. In the second class, the search direction is given by

$$d_k = \begin{cases} -g_k & \text{if } k = 0, \\ -g_k + \beta_k d_{k-1} & k \ge 1, \end{cases}$$

where the parameter β_k is a scalar known as the conjugate gradient parameter and g_k is the gradient of f. In this case, the iterative scheme for solving (1.1) corresponds to conjugate gradient method, it is one of the famous iterative scheme in this aspect due to its simplicity, low memory requirement and a good convergences property.

1.2 Basic Definitions and Theorems

We begin with the definitions and Theorems that are needed in this thesis which are in Sun and Yuan (2006)

Definition 1.1 The search direction d_k is said to be a descent direction of the objective function f, if $g_k^T d_k < 0$. Similarly, the search direction d_k is said to be sufficient descent direction of the objective function f if

$$g_k^T d_k \le -c \|g_k\|^2, \tag{1.5}$$

where c > 0.

Definition 1.2 A point $x^* \in \Re^n$ is said to be a stationary point (or a critical point) point of the objective function if $\nabla f(x^*) = 0$.

Definition 1.3 A point $x^* \in \Re^n$ is said to be a local minimizer of the objective function f, if $f(x^*) \leq f(x_k)$ for all $x_k \in \Re^n$.

Definition 1.4 A point $x^* \in \Re$ is said to be global minimizer of the objective function f, if $f(x^*) \leq f(x_k)$ for all $x_k \in \Re^n$ and $x_k \neq x^*$.

Definition 1.5 A point $x^* \in \Re^n$ is said to be a strict minimizer of the objective function f, if $f(x^*) < f(x_k)$ for all $x_k \in \Re^n$.

Lemma 1.1 Let U and V be matrices belonging to $\Re^{n \times m}$ for some $m \in [1, n]$ and suppose that $L(\Re^n)$ defined the linear space of all matrices of order n. The rank-k update of matrix that is nonsingular $A \in L(\Re^n)$ of the form

$$\hat{A} = A + UV^T$$

is nonsingular if and only if $\kappa = I + V^T A^{-1} U \neq \mathbf{0}$ specifically if $\kappa \neq \mathbf{0}$, then, the inverse of the matrix A is given by

$$\hat{A}^{-1} = A^{-1} - A^{-1}U(I + V^T A^{-1}U)^{-1}V^T A^{-1}, \qquad (1.6)$$

which is known as Sharman-Morrison formular (see Nocedal and Wright (2006)).

Theorem 1.1 Let $P \in \Re^n$ be a non empty open convex set and let $f : P \in \Re^n \to \Re$ be a differentiable function, then f is convex if and only if

$$f(y) \ge f(x) + \nabla f(x)^T (y - x)$$

for all $x, y \in P$. Similarly, f is strictly convex on P if and only if

$$f(y) > f(x) + \nabla f(x)^T (y - x)$$

for all $x, y \in P$. Further more f is strongly or uniformly convex if and only if

$$f(y) \ge f(x) + \nabla f(x)^T (y - x) + \frac{1}{2} c \|y - x\|^2$$

for all $x, y \in P$ where c > 0 is a constant.

Proof omitted see (Sun and Yuan (2006))

1.3 Optimality Conditions

The following Theorem (1.2-1.4) are due to (Sun and Yuan (2006))

Theorem 1.2 (First Order Necessary Condition) Let $f : U \in \mathbb{R}^n \to \mathbb{R}$ be continuously differentiable on an open set U, if $x^* \in \mathbb{R}^n$ is a local minimizer then $\nabla f(x^*) = 0$.

Theorem 1.3 (Second Order Necessary Condition) Let $f : U \in \Re^n \to R$ be continuously differentiable on an open set U, if $x^* \in U$ is a local minimizer then $\nabla f(x^*) = 0$ and $\nabla^2 f(x^*)$ is positive -semi definite.

Theorem 1.4 (Second Order Sufficient Condition) Let $U \in \Re^n \to \Re$ be continuously differentiable on an open set U, if $x^* \in U$ is a strict local minimizer then $\nabla f(x^*) = 0$ and $\nabla^2 f(x^*)$ is positive definite.

1.4 Rate of Convergence

The rate of convergence refers to the speed in which the sequence generated by an algorithm tends to its limit point (Nocedal and Wright (2006). The following definition (1.6-1.4) are from (Nocedal and Wright (2006) **Definition 1.6** Let the sequence $\{x_k\}$ in \Re^n converge to a finite limit x^* .

• The sequence is said to be q-linear convergence if there exist a scalar $r \in (0, 1)$ such that

$$\frac{\|x_{k+1} - x_*\|}{\|x_k - x_*\|} \le r$$

for sufficiently large k.

• The convergence is said also to be q-superlinear if

$$\lim_{k \to \infty} \frac{\|x_{k+1} - x_*\|}{\|x_k - x_*\|} = 0$$

• The convergence is said to be q-quadratic if there exist a scalar N such that

$$\frac{\|x_{k+1} - x_*\|}{\|x_k - x_*\|^2} \le N$$

for a very large k.

Definition 1.7 Let the sequence $\{x_k\}$ in \Re^n converge to a finite limit x^* . The sequence convergence is said to be r-(linear, superlinearly quadratically) if there exist a sequence converging q-(linear, superlinearly quadratically) to zero.

1.5 Motivation

Among the quasi-Newton methods is the DFP method which serve as one of the clever optimization method for determining the inverse Hessian approximation, its encompasses the merits of Newton's method and steepest descent method while avoiding their disadvantages Goldfarb (1969). It been neglected over the years as the result of

- 1. Computation and storage of the Hessian or inverse Hessian approximation in every iteration
- 2. Its inability to adjust within a fewer step whenever a less informative inverse Hessian matrix is approximated (Nocedal and Wright (2006)).

On the other hand, the conjugate gradient methods are fascinating iterative scheme with low memory requirement. For non-linear problems most conjugate gradient method cannot guarantee the sufficient descent condition

$$g_k^T d_k \le -c \left\| g_k \right\|$$

which facilitate it convergence (Gilbert and Nocedal (1992)).

1.6 Objectives of the Thesis

The main objectives are

- 1. To develop some three-term conjugate gradient method via DFP update which would satisfies both the sufficient and the conjugacy conditions
- 2. To establish the global convergence properties of the proposed methods
- 3. To present some numerical results by comparing the proposed method with other existing conjugate gradient methods.
- 4. To apply the proposed method on real-life optimization problems.

1.7 Structure of the Thesis

The remaining part of this thesis is structure as: An overview of nonlinear unconstrained optimization methods is present in chapter 2. Chapter 3 deals with the derivation of the proposed method STCG, the convergence properties, numerical result and discussion. In chapter 4 we present the real-life application of the proposed method on statistical problems. Chapter 5 present the implementation of the proposed method with nonmonotone line search and chapter 6 gives conclusion and future research.



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