

APG3013F Assignment 3 - Numerical methods
and quality control for least squares traverse
adjustments

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1 Introduction

The aim of this assignment is to solve a traverse least squares adjustment by Cholesky decomposition, and to perform a sequential least squares adjustment.

2 Background

2.1 Cholesky Decomposition

There are many techniques which can be used in when performing matrix operations on a large scale to reduce the cost of computation, and to speed up computation. For example, with calculations involving sparse matrices it is useful to partition matrices such that maximum multiplication of small numbers/elements/zeros occur. Matrix decomposition is another method of matrix operation optimization. Matrix decomposition essentially involves breaking down a matrix into a formulation of multiple matrices (more than one). Some other techniques are Gauss Reduction, Jacobi Method, Gauss-Jordan Method etc. The Cholesky Method is very convenient in that it provides an inverse which is later used for subsequent least squares matrices. It is for this reason that the Cholesky method is of particular suitability when programming a least squares solution.

2.2 Sequential least squares

Sequential least squares is a method least squares determination whereby additional observations are added to a calculation post computation (after an iteration). This can be very useful when filtering or smoothing is required (Kalman filter).

3 Problem Statement

3.1 Cholesky decomposition

Calculate a traverse solution by Cholesky decomposition when determining the x-vector of a least squares solution.

3.2 Sequential Least Squares

Sequentially add observations to an adjustment (Sequential least squares).

4 Method

The Cholesky decomposition is used as follows when solving for a least squares solution vector, x :

$$(A^t P A)x = A^t P l$$

5 Results

When performing a least squares solution using a program, solving for all resultant matrices, Cholesky decomposition results in the ability to solve for these resultant matrices much more quickly and more efficiently.

6 Discussion

7 Conclusion