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 by adding additional observations during iteration of a solution.

2 Background

2.1 Cholesky Decomposition

There are many techniques which can be used in when performing matrix opertaions on a large scale to reduce the cost of computation, and to speed up computation. For example, with calculations involving sparse matricies it is useful to partition matricies such that maximum multiplication of small numbers/elements/zeros occur. Matrix decomposition is another method of matrix opertaion optomization. Matrix decomposition essentially involves breaking down a matrix into a formulation of multiple matricies (more than one). Some other techniques are Guass Reduction, Jacobi Method, Gauss-Jordan Method etc. The Cholesky Method is very conveniant in that it provides an inverse which is later used for subsequent least squares matricies. It is for this reason that the Cholesky method is of partiular 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 matricies, Cholesky decomposition results in the ability to solve for these resultant matricies much more quickly and more efficiently.

6 Conclusion

6.1 Cholesky Decomposition

The Cholesky method of solving for the x vector in a least squares adjustment facilitates efficiency computation and can result in quicker computations.

6.2 Sequential Least Squares

Sequential least squares is very useful for smoothing and filtering applications, and is rather trivial to implement ontop of an existing least squares program.