Scalable Analysis Methods for Sparse Large-scale Systems

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Can Systems be Certified Distributively?



Componentwise performance verification without global model?

Outline

- Introduction
- Distributed Positive Test for Matrices
- Distributed Nonconservative System Verification
- A Scalable Robustness Test

The sparse matrix on the left is positive semi-definite if and only if it can be written as a sum of positive semi-definite matrices with the structure on the right.



Proof idea

The decomposition follows immediately from the band structure of the Cholesky factors:



[Martin and Wilkinson, 1965]

Example

The simplest decomposition is to just split each coefficient equally between the squares where it belong. This could work if the matrix is diagonally dominant:



Generalization

Cholesky factors inherit the sparsity structure of the symmetric matrix if and only if the sparsity pattern corresponds to a "chordal" graph.



[Blair & Peyton, An introduction to chordal graphs and clique trees, 1992]

Example: Non-chordal graph



Example: Chordal graphs

If T is a tree, then T^k is chordal for every $k \ge 1$.



A Theorem on Positive Extensions

A matrix with entries specified according to a chordal graph has a positive definite completion if and only if all fully specified principal minors are positive definite. [Grone, et.al, 1984]



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A System with Tridiagonal Structure



$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \vdots \\ \dot{x}_n(t) \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & & 0 \\ a_{21} & a_{22} & \ddots & \\ & \ddots & & a_{(n-1)n} \\ 0 & & a_{n(n-1)} & a_{nn} \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ \vdots \\ x_n(t) \end{bmatrix} + \begin{bmatrix} w_1(t) \\ w_2(t) \\ \vdots \\ w_n(t) \end{bmatrix}$$

A Sparse Stability Test

For the sparse matrix A, let the left hand side illustrate the structure of $(sI - A)^*(sI - A)$. Then the matrix is stable if and only if the right hand side split can be done with all squares positive definite for s in the right half plane.



Hence global stability can always be verified by local tests!

A Sparse Gain Bound

Solutions to
$$\dot{x}(t) = Ax(t) + w(t)$$
, $x(0) = 0$ satisfy

$$\int_0^T |x(t)|^2 dt \le \gamma^2 \int_0^T |w(t)|^2 dt$$

if and only if



where the terms on the right hand side are positive definite for s in the right half plane.

A Sparse Passivity Test

Suppose

$$\dot{x} = Ax + Bx + w \qquad x(0) = 0$$

$$y = Cx$$

Then

$$\int_0^T \left(\gamma^2 u(t) y(t) + |w(t)|^2 \right) dt \ge 0 \qquad \text{ for all } u, w, T$$

if and only if the matrix

$$\begin{bmatrix} (sI-A)^*(sI-A) & \gamma^2 C^T - (sI-A)^* B \\ \gamma^2 C - B^*(sI-A) & B^T B \end{bmatrix}$$

is positive semi-definite for $\operatorname{Re} s \ge 0$.

Passivity can be tested componentwise without conservatism!

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Robustness Analysis for Chained System



Many robustness analysis problems can be reduced to proving that $(I - \Delta(s)G(s))^{-1}$ is stable for $\Delta = \text{diag}\{\delta_1, \ldots, \delta_m\}$ with $|\delta_i(i\omega)| \le 1$. This can be done by finding $X(\omega) = \text{diag}\{x_1(\omega), \ldots, x_m(\omega)\} \succ 0$ with $X(\omega) \succ G(i\omega)X(\omega)G(i\omega)^*$ where

$$G(i\omega) = \begin{bmatrix} g_1 & h_1 & 0 & 0 & 0 \\ f_1 & g_2 & h_2 & 0 & 0 \\ 0 & f_2 & \ddots & \ddots & 0 \\ 0 & 0 & \ddots & g_{m-1} & h_{m-1} \\ 0 & 0 & 0 & f_{m-1} & g_m \end{bmatrix}$$

Note that each x_i influences at most nine elements of $X - GXG^*$.

Scalable Distributed Computations



The matrix $GXG^* - X$ is negative definite if and only if there exist y_i, z_i, w_i such that the following are negative definite:

$$W_{1} = \begin{bmatrix} g_{1} & h_{1} \\ f_{1} & g_{2} \\ 0 & f_{2} \end{bmatrix} \begin{bmatrix} x_{1} & 0 \\ 0 & x_{2} \end{bmatrix} \begin{bmatrix} g_{1} & h_{1} \\ f_{1} & g_{2} \\ 0 & f_{2} \end{bmatrix}^{*} - \begin{bmatrix} x_{1} & 0 & 0 \\ 0 & x_{2} + w_{1} & y_{1} \\ 0 & y_{1}^{*} & z_{1} \end{bmatrix}$$
$$W_{2} = \begin{bmatrix} h_{2} \\ g_{3} \\ f_{3} \end{bmatrix} x_{3} \begin{bmatrix} h_{2} \\ g_{3} \\ f_{3} \end{bmatrix}^{*} + \begin{bmatrix} w_{1} & y_{1} & 0 \\ y_{1}^{*} & z_{1} - w_{2} - x_{3} & -y_{2} \\ 0 & -y_{2}^{*} & -z_{2} \end{bmatrix}$$
$$\vdots$$
$$W_{m-2} = \dots$$

. . .

Conclusions

- Distributed Positive Test for Matrices
- Distributed Nonconservative System Verification
- Scalable Robustness Tests for Heterogeneous Systems