# Modeling Physics with Differential-Algebraic Equations 

## Lecture 2

Structural Analysis: Index Reduction

Master Cyber-Physical Systems (M2)

Khalil Ghorbal<br>khalil.ghorbal@inria.fr

## Summary lecture 1

(1) Ordinary Differential Equations:

- Cauchy-Lipschitz theorem: existence and uniqueness of solutions
- Liouville theorem: no closed form solutions in general
- Numerical integration: convergence and stability
- Qualitative analysis: invariant regions
(2) Differential-Algebraic Equations
- Informal Introduction
- Examples
- Different Forms


## Partial Solving of Implicit Systems

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Problem
Knowing (the state) \(X\), we would like to find \(Y\) sich that \(F(X, Y)=0\).
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## Implicit Function Theorem

Let $F(X, Y)$ : $n$ equations, where $|Y|=n,|X|=m$.
If $(u, v) \in \mathbb{R}^{m+n}$ is such that $F(u, v)=0$ and $J=\frac{\partial F}{\partial Y}$ is nonsingular at $(u, v)$, then there exists, in an open neighborhood $U$ pf $u$, a unique set of functions $G$ such that $v=G(u)$ and $F(w, G(w))=0$ for all $w \in U$.

## Structural Analysis

How to solve the scalability problem for large systems? At which cost?

## Outline

(1) Matching Problem

(2) BLT Decomposition

(3) Pantelides Algorithm

## Structural Analysis of Systems of Equations: Example

Consider the following system of equations:

$$
\left\{\begin{array}{l}
f_{1}\left(x_{1}\right)=0 \\
f_{2}\left(x_{1}, x_{2}, x_{3}\right)=0 \\
f_{3}\left(x_{3}\right)=0
\end{array}\right.
$$

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\end{array}\right.
$$

## Incidence Matrix

$f_{1}$
$f_{2}$
$f_{3}$$\left(\begin{array}{ccc}x_{1} & x_{2} & x_{3} \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 0 & 0 & 1\end{array}\right)$

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Bipartite Graph (Bigraph)


## Bigraphs and Matching

Bipartite graph ( $F, V, E$ )

- $F$ : set of equations
- $V$ : set of variables (disjoint with $F$ )
- $E$ : subset of the cartesian product $F \times V$

Example: a triangle is not a bipartite graph.

Matching Problem
Given a bipartite graph $(F, V, E)$, assign one and only one equation $f \in F$ to each variable $v \in V$ such that $(f, v) \in E$.

## Matching Problem: Intuitions

## Algorithm

For each equation, pick up its first unmatched variable. The procedure succeeds whenever all variables are matched.


## Matching Problem: Intuitions

## Algorithm (incomplete/wrong)

For each equation, pick up its first unmatched variable. The procedure succeeds whenever all variables are matched.


Backtrack $\left(f_{2}, x_{3}\right)$


## Matching Problem: Algorithm

Let $G:(V, E)$ denote a graph

- Matching: A matching is a set of pairwise non-adjacent edges.
- Maximum Matching: A matching having the maximum cardinality, denoted $\nu(G)$.
- Perfect Matching: A matching such that every vertex of the graph is incident to exactly one edge of the matching.

Proposition $G$ has a perfect matching if and only if $|G|=2 \nu(G)$.
$\rightarrow$ For systems of $n$ variables and $n$ equations: (1) compute $\nu(G)$, (2) check if $\nu(G)=n$.

## Hopcroft-Karp Algorithm (1973)

- Input: bipartite graph $(U, V, E)$
- Output: maximum cardinality matching
- Complexity: (worst case) $O(|E| \sqrt{|V|})$


## Augmented Shortest Path

At each phase:

- BFS: alternate between $U$ and $V$ where one starts from an unmatched variable in $U$ and reaches an unmatched variable in $V$ while following a matched edge from $V$ to $U$ (this gives an augmented shortest path).
- DFS: selects one shortest path out of the many selected ones by the BFS.
- update the matching set

Example
Source: Wikipedia


Input Graph


After BFS
Iteration 1


After BFS Iteration 2

## Maximum Transversal Problem



Incidence Matrix

Maximum Transversal Problem Finding a permutation that places a maximum number of non-zero on the diagonal of a sparse matrix.

2 Very useful prior to decomposing the matrix using Gaussian elimination.

## Outline

## (1) Matching Problem

(2) BLT Decomposition

## (3) Pantelides Algorithm

## Block-Lower-Triangular (BLT) Form

Goal: Given a square matrix $M\left(m_{i j} \in\{0,1\}\right)$, find a permutation to put $M$ in a block lower triangular form.
$f_{1}$
$f_{2}$
$f_{3}$
$f_{4}$$\left(\begin{array}{cccc}x_{1} & x_{2} & x_{3} & x_{4} \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0\end{array}\right)$
$f_{3}$
$f_{1}$
$f_{4}$
$f_{2}$$\left(\begin{array}{cccc}x_{2} & x_{1} & x_{3} & x_{4} \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1\end{array}\right)$

## BLT Decomposition

Blocks of dimension $>1$ (called algebraic loops) are (numerically) solved by Gaussian elimination if linear or Newton methods if nonlinear.

Three main steps
(1) Find a (perfect) matching
(2) Construct a dependency graph
(3) Find strongly connected components (Tarjan's algorithm)

## BLT Decomposition: Example

 Perfect MatchingBipartite Graph (Bigraph)


Incidence Matrix
$\left.\begin{array}{l} \\ f_{1} \\ f_{2} \\ f_{3} \\ f_{4}\end{array} \begin{array}{cccc}x_{1} & x_{2} & x_{3} & x_{4} \\ \hline 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & \boxed{1} \\ 0 & \boxed{1} & 0 & 0 \\ 1 & 1 & \boxed{1} & 0\end{array}\right)$

## BLT Decomposition: Example Dependency Graph



Dependency Graph


# BLT Decomposition: Example 

Strongly Connected Components



## BLT Decomposition

$f_{3}$
$f_{1}$
$f_{4}$
$f_{2}$$\left(\begin{array}{cccc}x_{2} & x_{1} & x_{3} & x_{4} \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1\end{array}\right)$

## Strongly Connected Components (SCC)

## Definitions

- A directed graph is strongly connected if every vertex is reachable from every other vertex.
- Strongly connected components of a directed graph form a partition into subgraphs which are themselves strongly connected.
- By contracting each strongly connected component into one vertex, one obtains a condensation of the original graph into a directed acyclic graph.

Tarjan's Algorithm (1972)

- One depth-first search (vs. 2 for Kosaraju's algorithm (1978))
- Simple and elegant data structure
- Complexity: $O(|V|+|E|)$


## Structural Analysis

## Structural NonSingularity

A square matrix is said to be structurally nonsingular if it remains almost everywhere nonsingular when its nonzero coefficients vary over some neighborhood.

Relation to the BLT Decomposition
The Jacobian $J=\frac{\partial F}{\partial Y}$ is structurally nonsingular if and only if $G_{F}$ (the incidence graph related to $F$ ) can be decomposed in a BLT form.

## Outline

## (1) Matching Problem

(2) BLT Decomposition
(3) Pantelides Algorithm

## Differential Algebraic Equations

General Form

$$
F(x, \dot{x}, y, t)=0
$$

- state variables: $x \in \mathbb{R}^{n}$
- algebraic variables: $y \in \mathbb{R}^{m}$
- $F: G \subset \mathbb{R}^{n} \times \mathbb{R}^{n} \times \mathbb{R}^{m} \times \mathbb{R} \rightarrow \mathbb{R}^{m+n}$

Index
The index of an DAE is the minimum number of times that all or part of the DAE must be differentiated with respect to time $(t)$ in order to determine $\dot{x}$ as a continuous function of $x$ and $t$.
(Brenan, Campbell 1996)

# Example: Pendulum 



System of Equations
$f_{1}: \dot{x}=u$
$f_{2}: \dot{y}=v$
$f_{3}: \dot{u}=-\lambda x$
$f_{4}: \quad \dot{v}=-\lambda y-g$
$f_{5}: \quad 0=L^{2}-x^{2}-y^{2}$

Incidence Matrix
$f_{1}$
$f_{2}$
$f_{3}$
$f_{4}$
$f_{5}$$\left(\begin{array}{ccccc}\dot{x} & \dot{y} & \dot{u} & \dot{v} & \lambda \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0\end{array}\right)$

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No perfect matching

# Example: Pendulum 

Exhibiting Latent Equations
Suppose a consistent initialization

- $f_{5}$ is not used
- $f_{5}$ holds for all $t$, then $\dot{f}_{5}$ has to hold for all $t$
- $\dot{f}_{5}: 2 x \dot{x}+2 y \dot{y}=0$ (symbolic differentiation)

Incidence Matrix
$f_{1}$
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# Example: Pendulum 

Exhibiting Latent Equations
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Still no matching !

Bipartite Graph


# Example: Pendulum 

Higher Indices

Incidence Matrix
$f_{1}\left(\begin{array}{ccccccc}\dot{x} & \dot{y} & \dot{u} & \dot{v} & \lambda & \ddot{x} & \ddot{y} \\ f_{2} \\ f_{3} \\ f_{4} \\ \ddot{f}_{5} \\ \ddot{f}_{5} \\ \dot{f}_{1} \\ \dot{f}_{2}\end{array}\left(\begin{array}{lllllll}0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1\end{array}\right)\right.$

Matching Successful!

Bipartite Graph


## Pantelides Algorithm

Pantelides algorithm (1988) attempts to decompose the function $F$ of a given DAE into a BLT form by exhibiting latent equations.

- First structural analysis of DAE
- Not guaranteed to terminate
- Applies only to first-order systems
- May overestimate the differential index
- Other methods: Signature ( $\Sigma$ ) method, J. D. Pryce (2001)


## Structural Analysis (Complement)

Consider the following DAE

$$
\begin{aligned}
& \dot{z}-\dot{x} y-x \dot{y}+2 x+y-3=0 \\
& z-x y=0 \\
& x+y-2=0
\end{aligned}
$$

- Pantelides: structural index 1
- Differentiation index is 0 (simple linear system)
- (Hidden) Cancellation problems are undecidable in general


## To summarize

## Index Reduction

- Given a DAE $F(x, \dot{x}, t)$, we have seen how to perform a structural analysis to numerically compute $\dot{x}$ function of $x$. The structural nonsingularity ensures that, generically, one can perform the computation following the block order suggested by the BLT decomposition. Thus, one is able to compute the numerical values of the derivatives given a consistent state of the system and carry on with a standard numerical integration.
- Note: the structural index is not always equal to the differentiation index (see the first reference below for concrete examples).


## References

- Guangning Tan, Ned S. Nedialkov, John D. Pryce, Symbolic-Numeric Methods for Improving Structural Analysis of Differential-Algebraic Equation Systems, Book Chapter in "Mathematical and Computational Approaches in Advancing Modern Science and Engineering", pp. 763-773, Springer, 2016 (available on arXiv:1505.03445 [cs.SC])
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- Rajeev Motwani, Average-case Analysis of Algorithms for Matchings and Related Problems, Journal of the ACM, 41 (6): pp. 1329-1356 (1994)
- Tarjan, R. E., Depth-first search and linear graph algorithms, SIAM Journal on Computing, 1 (2): pp. 146-160 (1972)

