Schedulability Analysis for Systems with Data and Control Dependencies

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Outline

- Motivation
- System Model
- Problem Formulation
- Schedulability Analysis
- Experimental Results
- Conclusions





Performance estimation:

Based on schedulability analysis.

Schedulability analysis:

- Worst case response time of each process.
- Models in the literature:
 - Independent processes;
 - Data dependencies: release jitter, offsets, phases;
 - Control dependencies: modes, periods, recurring tasks.



Characteristics and Message

Characteristics:

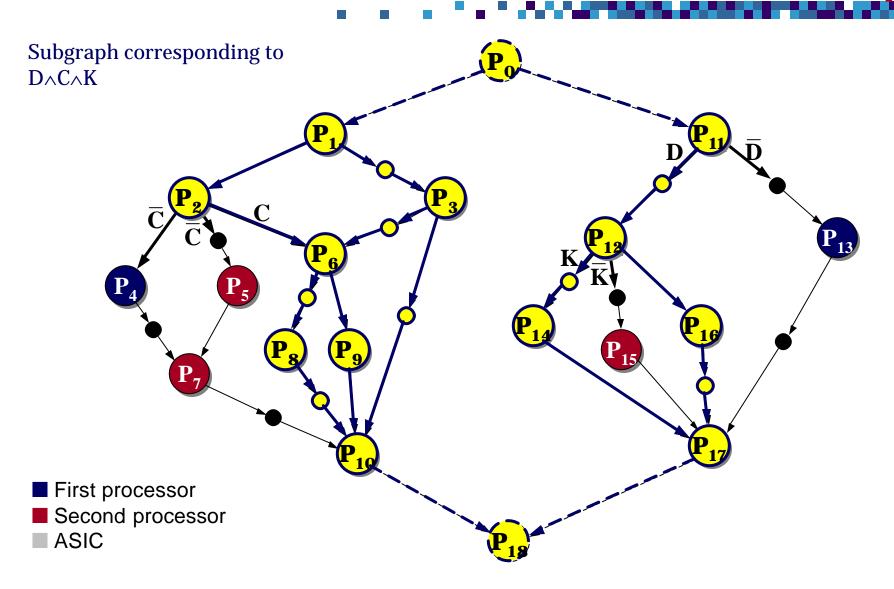
- Heterogeneous system architecture.
- Fixed priority preemptive scheduling.
- Systems with data and control dependencies.
- Tighter worst case delay estimations.

Message:

The pessimism of the analysis can be drastically reduced by considering the conditions during the analysis.



Conditional Process Graph





Problem Formulation

Input

- An application modelled as a set of conditional process graphs (CPG).
- Each CPG in the application has its own independent period.
- Each process has an execution time, a deadline, and a priority.
- The system architecture and mapping of processes are given.

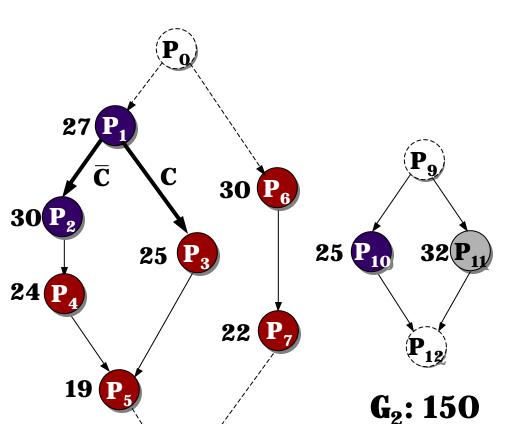
Output

Schedulability analysis for systems modelled as a set of conditional process graphs (both data and control dependencies).

- Fixed priority preemptive scheduling.
- Communication of messages not considered, but can be easily added.



Example



CPG	Worst Case Delays	
	No conditions	Conditions
G ₁	120	100
\mathbf{G}_2	82	82

G₁: 200

Task Graphs with Data Dependencies

K. Tindell: Adding Time-Offsets to Schedulabilty Analysis, Research Report Offset: fixed interval in time between the arrival of sets of tasks. Can reduce the pessimism of the schedulability analysis.

Drawback: how to derive the offsets?

T. Yen, W. Wolf: Performance Estimation for Real-Time Distributed Embedded Systems, IEEE Transactions On Parallel and Distributed Systems Phase (similar concept to offsets).

Advantage: gives a framework to derive the phases.



Schedulability Analysis for Task Graphs

```
DelayEstimate(task graph G, system S)
        for each pair (P<sub>i</sub> P<sub>i</sub>) in G
                 \max \left[ P_{i} P_{j} \right] = \infty
        end for
                                                worst case response times and
        step = 0
                                                upper bounds for the offsets
        repeat
                 LatestTimes(G)
                 EarliestTimes(G)◀
                                                 lower bounds for the offsets
                 for each P_i \in G
                         MaxSeparations(P<sub>i</sub>)
                 end for
        until maxsep is not changed or step < limit
        return the worst case delay \delta_{_{\mathrm{G}}} of the graph G
end DelayEstimate
                                          maximum separation:
                                         maxsep[P_{i}, P_{j}]=0 if the execution of
                                          the two processes never overlaps
```



Schedulability Analysis for CPGs, 1

Two extreme solutions:

- Ignoring Conditions (IC)
 Ignore control dependencies and apply the schedulability analysis for the (unconditional) task graphs.
- Brute Force Algorithm (BF)

Apply the schedulability analysis after each of the CPGs in the application have been decomposed in their constituent unconditional subgraphs.

Schedulability Analysis for CPGs, 2

In between solutions:

■ Conditions Separation (CS)

Similar to *Ignoring Conditions* but uses the knowledge about the conditions in order to update the **maxsep** table: $maxsep[P_i, P_j] = 0$ if P_i and P_j are on different conditional paths.

■ Relaxed Tightness Analysis (two variants: RT1, RT2)

Similar to the *Brute Force Algorithm*, but tries to reduce the execution time by removing the iterative tightening loop (relaxed tightness) in the **DelayEstimation** function.

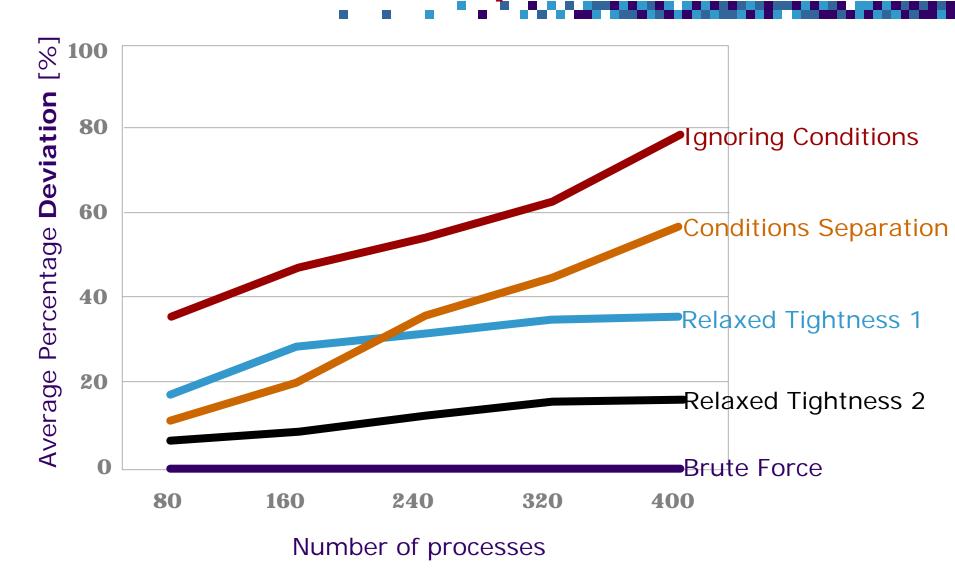
Experiments Setup

- Number of Graphs: 15030 for each dimension of 80, 160, 240, 320, 400 nodes;2, 4, 6, 8, 10 conditions.
- Graphs Structure:
 Random and regular (trees, groups of chains).
- Architecture:2, 4, 6, 8, 10 nodes.
- Mapping: 40 processes / node; random and using simple heuristics.
- Cost function: degree of schedulability

$$Cost \ function = \sum_{i=1}^{n} \left(D_{\Gamma_i} - \boldsymbol{d}_{\Gamma_i} \right)$$

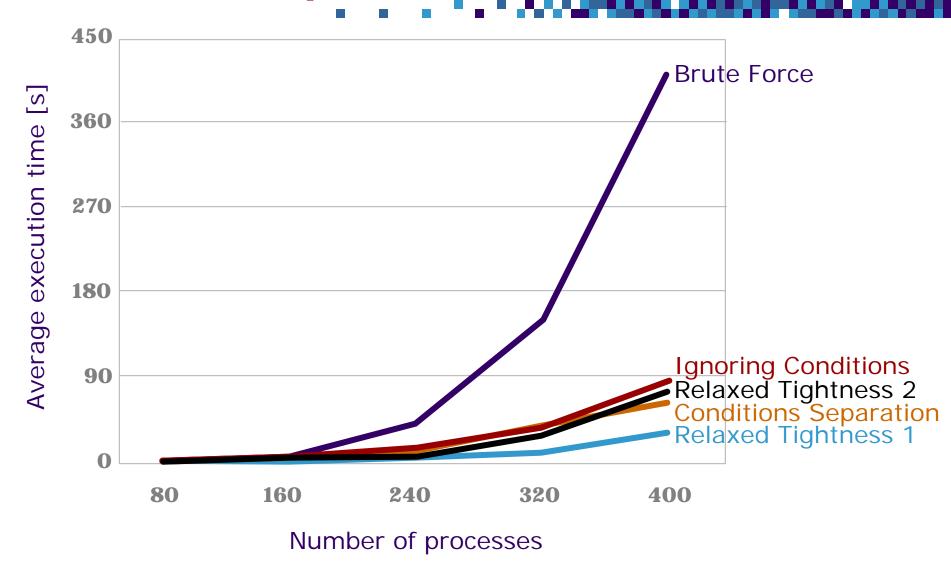


Experimental Results





Experimental Results (Cont.)





Real Life Example

- Vehicle cruise controller.
- Modelled with a CPG of 32 processes and two conditions.
- Mapped on 5 nodes: CEM, ABS, ETM, ECM, TCM.

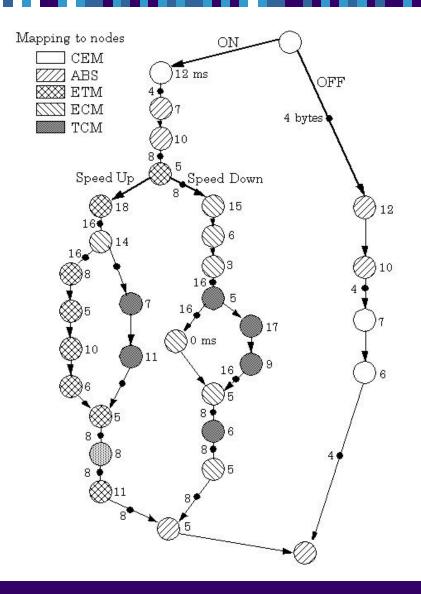
■ Deadline 130:

■ Ignoring Conditions: 138 ms

Conditions Separation: 132 ms

Relaxed Tightness 1, 2: 124 ms

■ Brute Force: 124 ms



Conclusions

- Schedulability analysis for hard real-time systems with control and data dependencies.
- The systems are modelled using conditional process graphs that are able to capture both the flow of data and that of control.
- Distributed architectures, fixed priority scheduling policy.
- Five approaches to the schedulability analysis of such systems.
- Extensive experiments and a real-life example show that: considering the conditions during the analysis the pessimism of the analysis can be significantly reduced.

