

Beyond the m-k model: restoring performance considerations in the time abstraction

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EMSOFT Tutorial - Seoul October, 2017

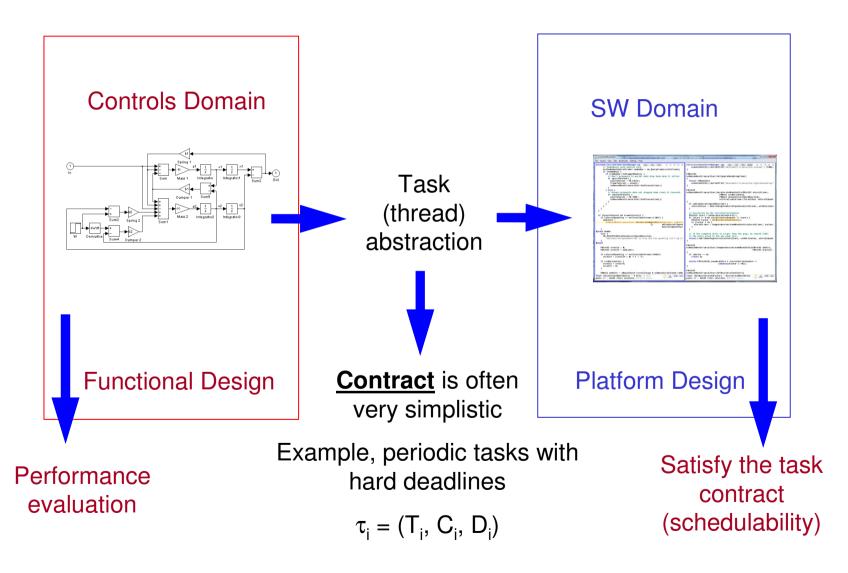


Summarizing...

- It is about the timing abstractions...
- But touches on a lot of issues
 - What are the available abstractions?
 - To what degree are timing abstractions possible/desirable at all?
 - What is the real consequence for the application of missing a deadline (depends on the program structure) ?
 - How do you manage a missed deadline (continue, drop...)?
 - Availability vs. Dependability should we drop the pretense of being fully predictable at design time?



Typical assumption for separations of concerns





- There is a very small number of "true" hard real-time systems
- Need for models that account for overload in several control domains (where deadline can be missed)
- Task models: from Hard to Firm to Soft deadline model to Value-based scheduling
 - Still not very satisfactory (declining trend in conference proceed.)
- The timing analysis community likes to build its own models (possibli inspired but drifting away from requirements)
 - Informally, we can miss deadlines but not too many and not too many in a row
 - (we don't get too much help from industry)
 - This is where the m-k model probably originated



The m-k model

- By now you heard plenty about it
- You can miss at most m deadlines out of k instances
- Not the only possible abstraction
 - Lenght and number of deadline misses in longest busy period
 - Other options ...



A summary of our work on m-k analysis

To be presented here (EMSOFT) on Tuesday

Y. Sun, M. Di Natale "Weakly Hard Schedulability Analysis for Fixed Priority Scheduling of Periodic Real-Time Tasks"

Contributions wrt previous work

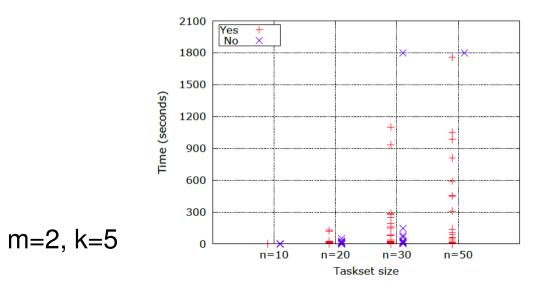
- Is not restricted to offset determined systems
- Can sweep a range of m-k options and even find the minimum m for a given k
 - Easily done since it is based on an optimization formulation

Limitations

- Still limited to periodic load
- Does not scale beyond 20 tasks and k>10
 - Do we really need those?



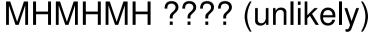
- How? Formulate the problem as a MILP
- Relaxing some constraints (number of interferences) and then refining to limit pessimism
- Feasibility or optimization formulation
 - Maximize # of misses m in any given window of k
- Results *very* close to true optimum
- Runtimes acceptable for many configurations





But you often can miss deadlines

- Problem with m-k model
- It is still a binary assessment (black and white)
 - No deadline miss→correct
 - Deadline miss →critical failure
 - Does not account for performance
- It is stateless (the position of the deadline misses in the sequence does not count)
 - is MMMHHH same as



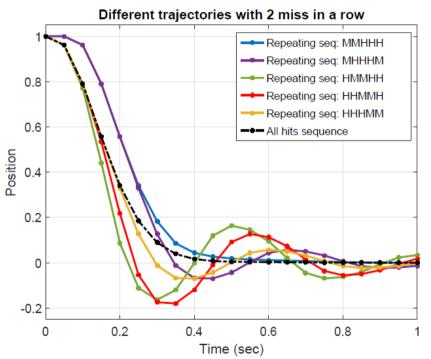


Fig. 1. Comparing different trajectories with the same constraint of deadline missed in a row. In general, changing the order of the missed deadlines in a H/M sequence leads to different behaviours, that cannot be discerned by the (h, n) description.



- We want to restore performance in the task contract
- How can you do that?
- It turns out it is not simple at all
 - And not because of a lack of fantasy in creating the contract...
- Depends on many assumptions and design choices

Also, consider the following observation

(M. Neukirchner @Waters workshop)

- "The real time community is very concerned about availability but should be more concerned about reliability"
- Support synthesis of monitors

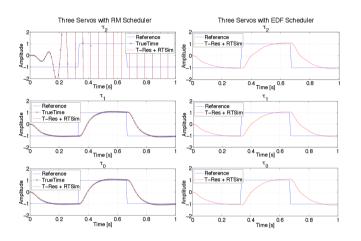


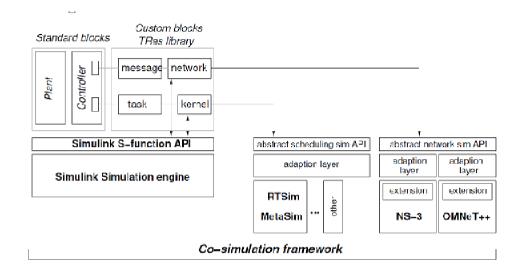
- Other alternative: avoid abstractions and perform the joint analysis (by model checking or (co)simulation) of the task and controls model
- By simulation Truetime, TRes (Simulink based), simulate the scheduler and the tasks together with the controls logic
- By formal models Hybrid systems, SpaceX

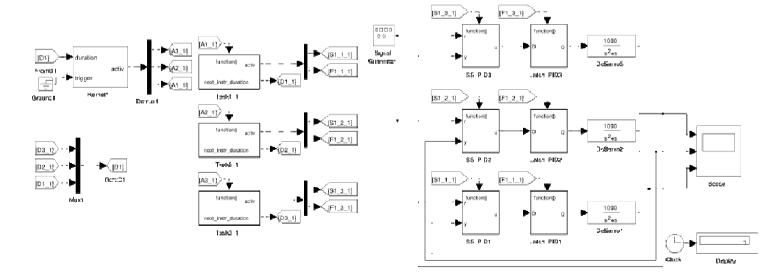


The T-Res approach

Co-simulation in Simulink







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The effect of a deadline miss heavily depends on the SW architecture

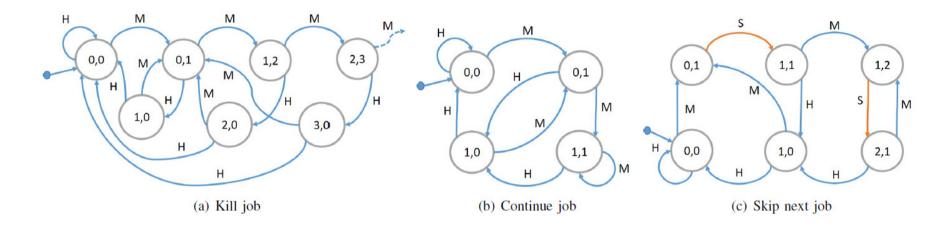
- It may be a delayed output (the task directly outputs)
- It may be an output at the usual time with old data (the task fails to update a TPU programming or misses a cycle in asynchronous cmmunication)
- It may be data that gets overwritten and completely missed



- When a task misses a deadline
 - Should you let it finish?
 - Accept the late termination, hope it is temporary
 - Should you terminate it at the deadline?
 spare load, give better chance to next instance to complete in time
 - Should you let it finish but skip the next instance?
 try to recover from overload
 - Should you terminate and skip?

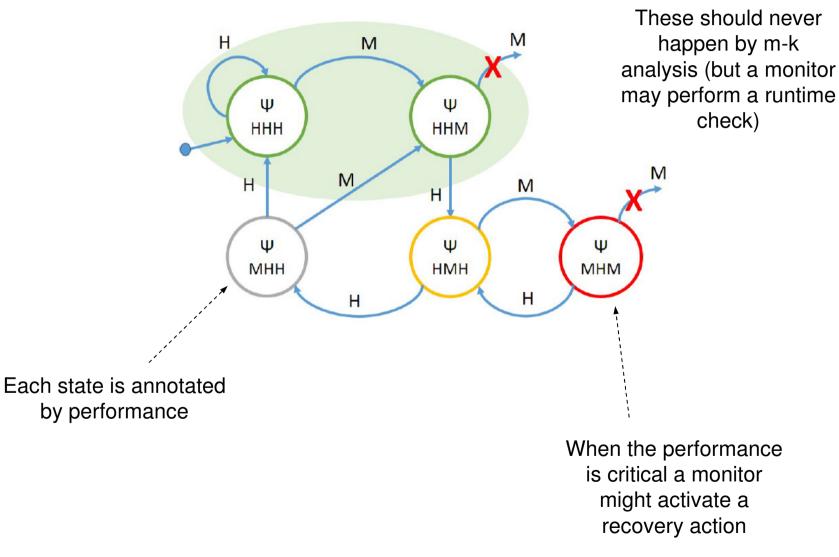


Effect of deadline misses



- *Hypothesis*: a Deadline miss results in use of old data by one cycle
- The delay of the output value depends on the management miss policy
- If you bound the possible sequences of Hits/Misses (by standard m-k analysis) you get a finite number of possible states for the data delays
- Each state can now be annotated with the corresponding performance

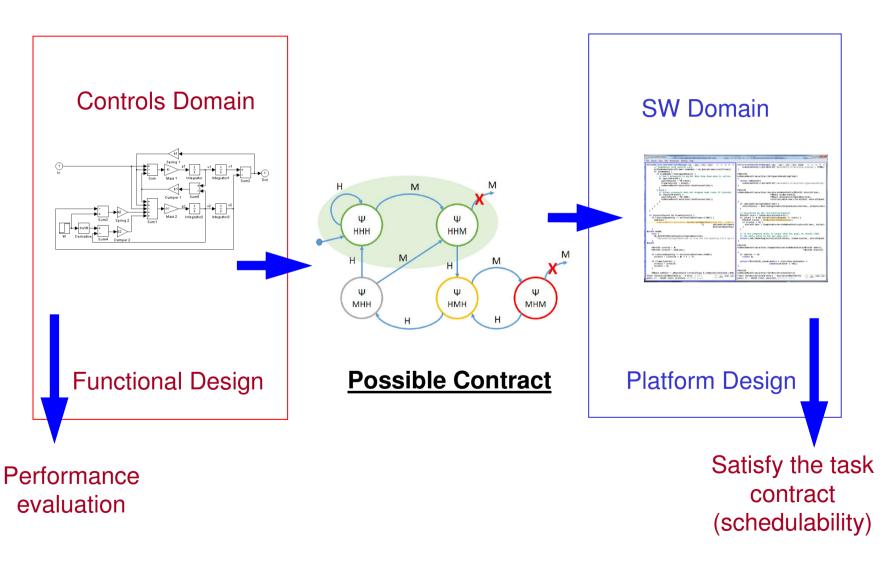




SOLA .



Typical assumption for separations of concerns





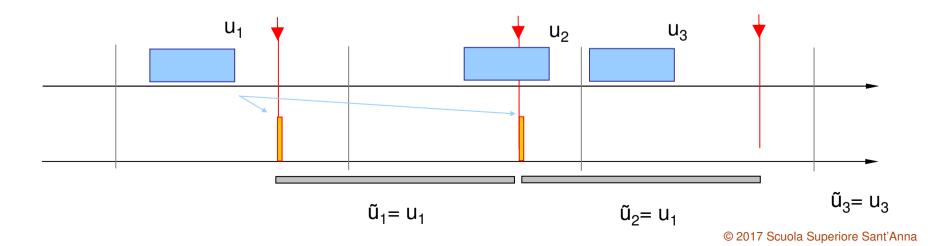
- How is this state machine computed?
- Analytically
 - For (simple?) LTI models
 - With a number of assumptions on sampling, actuation, deadline management ...
- Experimentally/by Simulation
 - Truetime, TRes ...



$$\dot{x}_c(t) = A_c x_c(t) + B_c u_c(t),$$

$$y_c(t) = C_c x_c(t),$$

- Controller implemented by τ_i period T_i , deadline D_i with $D_i \leq T_i$
- The job activated at time kT_i uses the state sensed at activation time kT_i (no sensing jitter). The actuator uses the output computed at the deadline ($kT_i + D_i$), and keeps it constant until ((k + 1) $T_i + D_i$).





$$\dot{x}_c(t) = A_c x_c(t) + B_c u_c(t),$$

$$y_c(t) = C_c x_c(t),$$

• The system is discretized as

$$x[k+1] = A_d x[k] + B_{d1} u[k-1] + B_{d2} u[k],$$

$$y[k] = C_d x[k]$$

• With Bd1 and Bd2 expressed by

$$A_{d} = e^{A_{c}T_{i}}, C_{d} = C_{c}$$
$$B_{d1} = \int_{T_{i}-D_{i}}^{T_{i}} e^{A_{c}s} ds B_{c}, B_{d2} = \int_{0}^{T_{i}-D_{i}} e^{A_{c}s} ds B_{c}$$



If no deadline miss

$$\tilde{u}_k = \begin{cases} u[k-1] & \text{for } t \in [kT_i, kT_i + D_i) \\ u[k] & \text{for } t \in [kT_i + D_i, (k+1)T_i) \end{cases}$$

If deadlines are missed

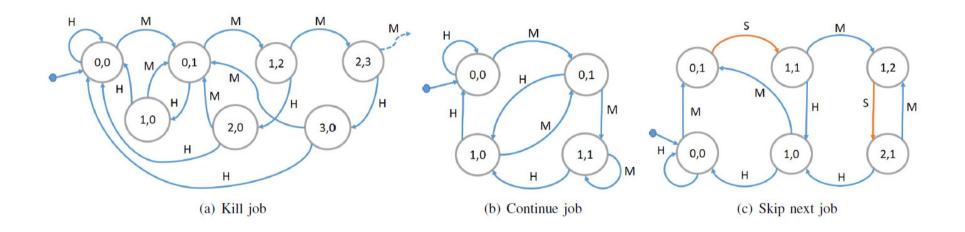
$$\tilde{u}_k = \begin{cases} u[k - \Delta_p - 1] & \text{for } t \in [kT_i, \ kT_i + D_i) \\ u[k - \Delta_c] & \text{for } t \in [kT_i + D_i, \ (k+1)T_i) \end{cases}$$

 Δ_p (previous) is the update age of the control output in the time interval [(k+1)T_i +D_i; kT_i +D_i)

 Δ_c (current) output age in the interval [kT_i + D_i; (k + 1)T_i + D_i). Need the pair Δ_p , Δ_c for all possible sequences of deadline hits and misses



Need the pair $\Delta_{\rm p},\,\Delta_{\rm c}$ for all possible sequences of deadline hits and misses





Expressing *u* as function of Δ_p , Δ_c in the state equation

$$x[k+1] = A_d x[k] + B_{d1} u[k-1-\Delta_p] + B_{d2} u[k-\Delta_c]$$

For a simple state feedback control

$$u[k] = K(r[k] - x[k])$$

If the reference r[k] is the null vector

$$x[k+1] = A_d x[k] - B_{d1} K x[k-1-\Delta_p] - B_{d2} K x[k-\Delta_c]$$



If x[k] is an extended state vector representing the state at the past Δ_{max} +1 steps

$$\mathbf{x}[k] = \left[x[k], x[k-1], \cdots, x[k-\Delta_{max}-1]\right]^T$$

We have

$$\mathbf{x}[k+1] = \begin{bmatrix} A_d & \cdots & -B_{d2}K & \cdots & -B_{d1}K & \cdots \\ \mathbf{I}_n & \mathbf{0}_n & \cdots & \cdots & \cdots & \cdots \\ \mathbf{0}_n & \mathbf{I}_n & \mathbf{0}_n & \cdots & \cdots & \cdots \\ \vdots & \cdots & \ddots & \ddots & \cdots & \cdots \end{bmatrix} \mathbf{x}[k]$$

The position of the terms B_{d1} and B_{d2} depends on Δ_{c} and Δ_{p}

$$\mathbf{x}[k+1] = \mathbf{\Phi}(\Delta_p, \Delta_c, \Delta_{max}) \mathbf{x}[k]$$



- Φ depends on the deadline miss management
- If jobs that miss deadlines are allowed to continue, for example ...

$$\Phi(0,0,1) = \begin{bmatrix} A_d - B_{d2}K & -B_{d1}K & \mathbf{0}_n \\ \mathbf{I}_n & \mathbf{0}_n & \mathbf{0}_n \end{bmatrix}$$

$$\Phi(0,1,1) = \begin{bmatrix} A_d & -BK & \mathbf{0}_n \\ \mathbf{I}_n & \mathbf{0}_n & \mathbf{0}_n \\ \mathbf{0}_n & \mathbf{I}_n & \mathbf{0}_n \end{bmatrix}$$

$$\Phi(1,0,1) = \begin{bmatrix} A_d - B_{d2}K & \mathbf{0}_n & -B_{d1}K \\ \mathbf{I}_n & \mathbf{0}_n & \mathbf{0}_n \\ \mathbf{0}_n & \mathbf{I}_n & \mathbf{0}_n \end{bmatrix}$$

$$\Phi(1,1,1) = \begin{bmatrix} A_d & -B_{d2}K & -B_{d1}K \\ \mathbf{I}_n & \mathbf{0}_n & \mathbf{0}_n \\ \mathbf{0}_n & \mathbf{I}_n & \mathbf{0}_n \end{bmatrix}$$



For the cumulative quadratic error index metric (for example)

$$P_{cqe} = \sum_{i=0}^{N} x[i]^T x[i].$$

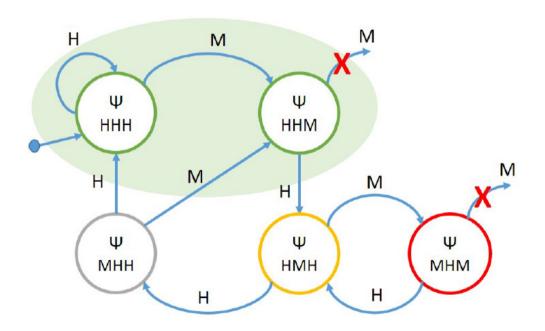
Assuming a given sequence of hits/misses

$$\begin{aligned} \mathbf{P}_{cqe}(s) &= \sum_{i=0}^{N} \mathbf{x}[i]^{T} \mathbf{x}[i] \\ &= \mathbf{x}_{0}^{T} \left(\mathbf{I} + \mathbf{\Phi}_{1}^{T} \mathbf{\Phi}_{1} + \mathbf{\Phi}_{1}^{T} \mathbf{\Phi}_{2}^{T} \mathbf{\Phi}_{2} \mathbf{\Phi}_{1} + \dots \right. \\ &+ \mathbf{\Phi}_{1}^{T} \mathbf{\Phi}_{2}^{T} \cdots \mathbf{\Phi}_{N-1}^{T} \mathbf{\Phi}_{N}^{T} \mathbf{\Phi}_{N} \mathbf{\Phi}_{N-1} \cdots \mathbf{\Phi}_{2} \mathbf{\Phi}_{1} \right) \mathbf{x}_{0} \\ &= \mathbf{x}_{0}^{T} \mathbf{\Psi}(s) \mathbf{x}_{0} \end{aligned}$$

The matrix $\Psi(s)$ can be computed as a function of the matrices $\Phi(\Delta_p, \Delta_c, \Delta_{max})$ generated by the sequence



Hence the performance annotation for the state model ...





Conclusion

Thank you!

