

Discrete and Continuous Optimal Control for Energy Minimization in Real-Time Systems

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25 September 2020

EBCCSP2020

EPFL



Introduction

- Minimizing the energy consumption J of an embedded system under real-time (RT) constraints.
- A single-core processor executes a finite set of jobs.
- HRTS: Hard Real Time System.
- HRT constraints: Each job must finish before its deadline.
- The processor can change its speed s , with DVFS technology.

$$\begin{array}{lcl} s \nearrow & \Rightarrow & \text{job execution time} \searrow \\ s \nearrow & \Rightarrow & J \nearrow \end{array}$$

Constrained Optimal Control Problem

Online Optimal Control Problem:

*At time t , the **decision maker (processor)** chooses the **speed** $s(t)$ based on the current state (**jobs released** before t) and statistical information about the **future job arrivals**.*

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***Minimize** the expected total energy cost while **satisfying** all the deadline constraints on the jobs.*

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- **Continuous** Time Control (CTC): at any time $t \in \mathbb{R}$ the controller chooses the current speed.

$\Rightarrow \nexists$ general algorithm.

- **Discrete** Time Control (DTC): decisions can only be made at the release times of the jobs.

$\Rightarrow \exists$ optimal solutions for this problem: **Dynamic Programming (DP)**

Our Solution to Solve the Continuous Time Control

- Proving that the continuous time control (CTC) solution is **not better than** the discrete time control (DTC) solution.
- As a consequence **any optimal DTC solution is also an optimal CTC solution.**

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Outline of the talk:

- ① Model description.
- ② Discrete Time Control (DTC) solution.
- ③ Continuous Time Control (CTC) solution.

Model Presentation

HRTS Model Presentation

- A **single uni-core** processor, equipped with DVFS, running at **speed** $s(t)$, with $s(t) \in \mathcal{S} = \{0, \dots, s_{\max}\}$.
- Finite set of **jobs** over a finite horizon T .

Jobs (r, c, d) characterized by :

- $r \in \mathbb{N}$: release time.
- $c \in \mathbb{N}$: **execution time (size)**.
- $d \in \mathbb{N} \mid d \leq \Delta$: **relative deadline**.

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- **Power consumption at speed s** : $Q(s(t))$

Q is a convex and increasing function in s .

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- **Energy**: $J = \int_0^T Q(s(t))dt$

Problem Statement

- ① What is the **scheduling policy** (j_π) of jobs to ensure that no job misses its deadline ?

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⇒ Choosing the processor speed $s(t)$ at each instant t

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Optimization under **constraint**:

- 1 Each job finishes before its deadline.
- 2 The expected energy consumption is minimized.

Scheduling Policy Choice

- What is the scheduling policy (j_π) to choose to be certain that each job is executed before its deadline ?

| | | |
|--|---|--|
| <ul style="list-style-type: none">• $\{s_\pi, j_\pi\}$ executes all jobs before their deadlines.• EDF policy is optimal for scheduling jobs. | } | $\Rightarrow \{s_\pi, j_{EDF}\}$ also executes all jobs before their deadlines. |
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\Rightarrow Choice of Earliest Deadline First (EDF)

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New Optimisation Problem

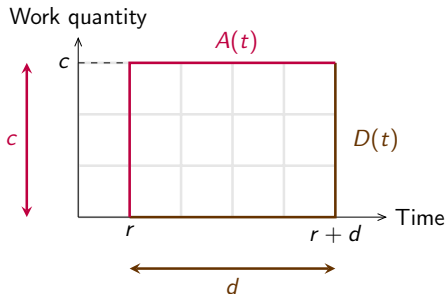
What is the optimal **speed policy** such that **the expected energy consumption is minimized** and **no job misses its deadline** ?

Discrete Time Control (DTC)

Work Quantity for Discrete Time Control

A single Job (r, c, d) :

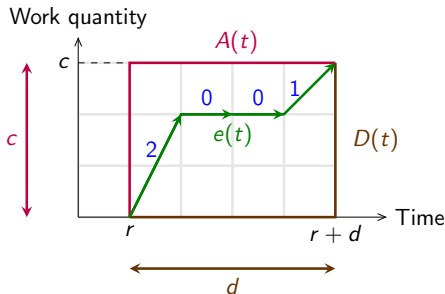
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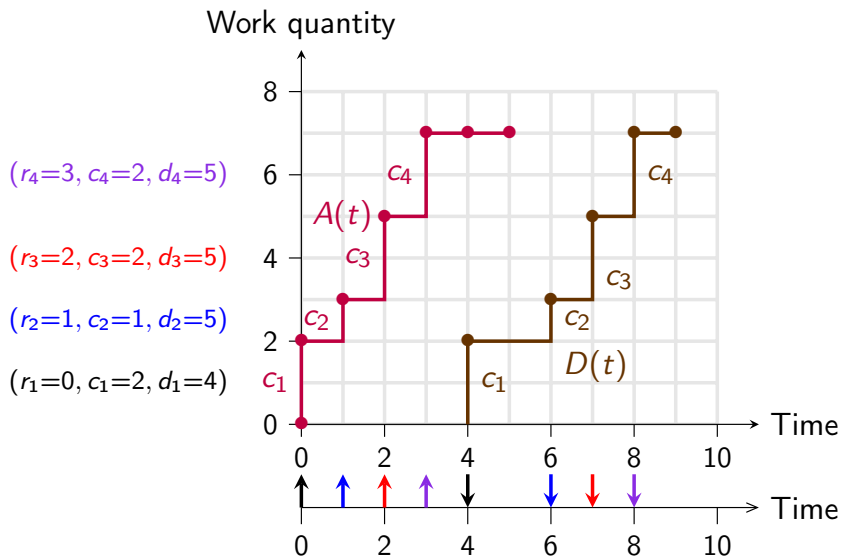
- $A(t)$ = work quantity arrived until time t .
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- $e(t)$ = work executed by the processor until time t with speeds $s(u)$.



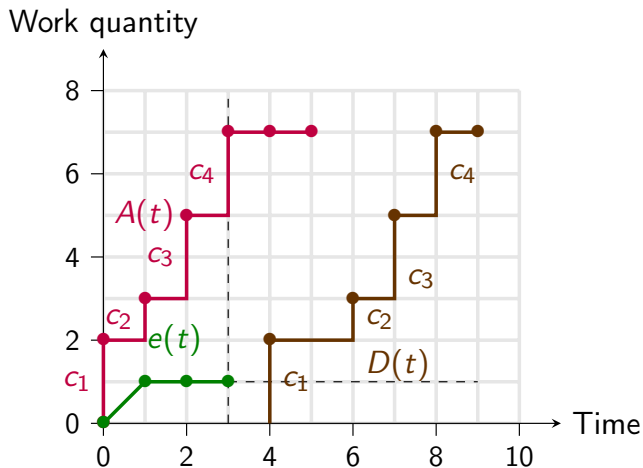
Two constraints on $e(t)$:

- 1 $e(t) \geq D(t)$: The job must be finished before its deadline.
- 2 $e(t) \leq A(t)$: The job can not be executed before it arrives.

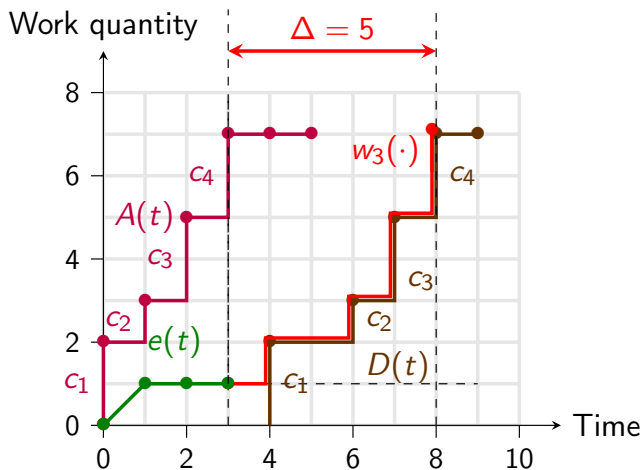
DTC: Remaining Work $w_t(\cdot)$ with $\Delta = 5$



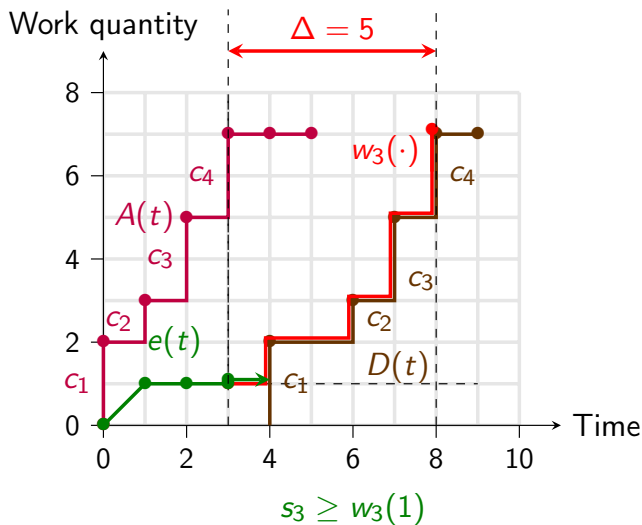
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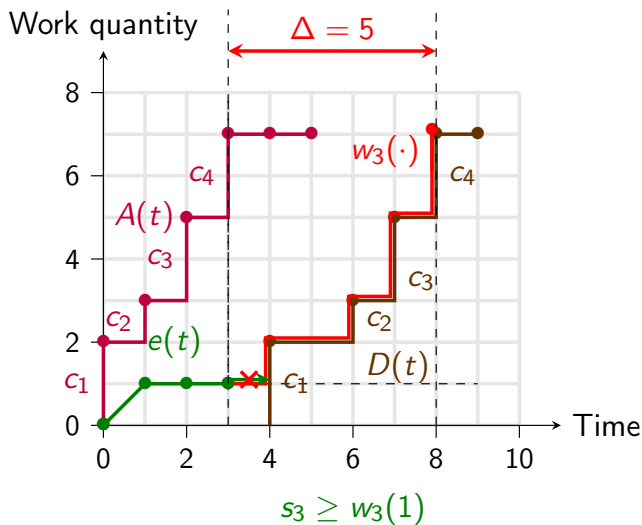
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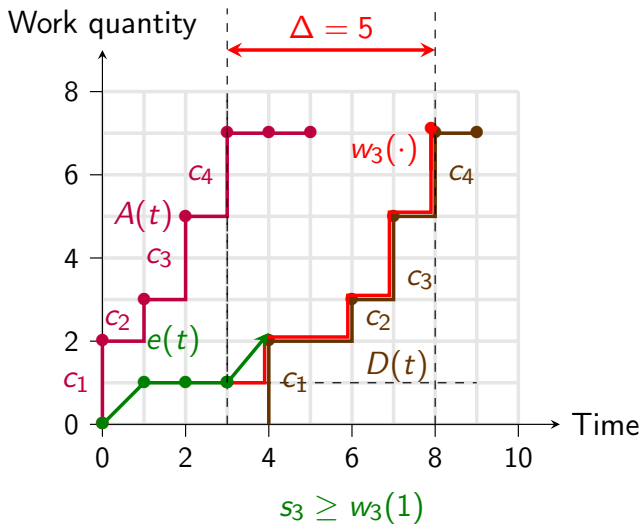
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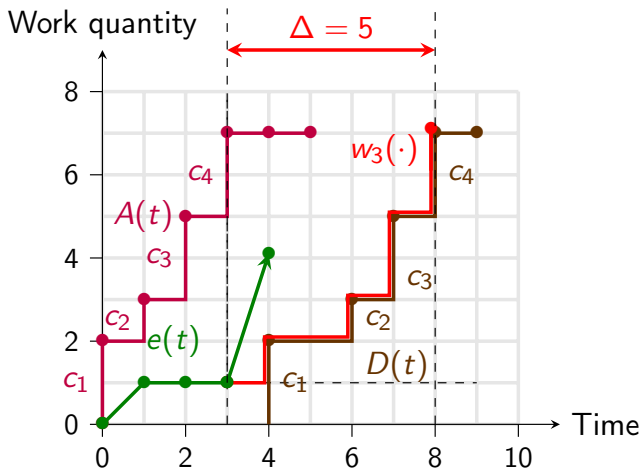
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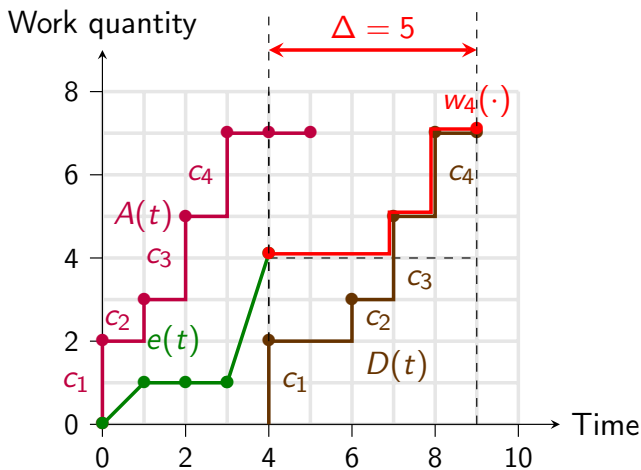
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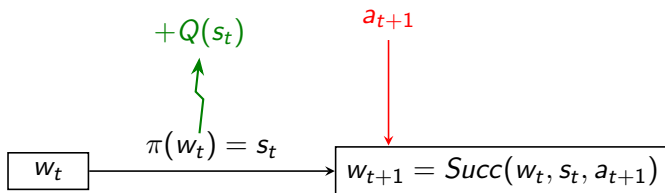
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DTC: Optimal Control under HRT Constraints

- a_t = Work quantity arrived at t .
- w_t = System state at t .
- s_t = Speed used by the processor at t .

State evolution under speed policy π :



With the HRT constraint: $s_t \geq w_t(1)$

DTC: Finite Horizon Markov Decision Process (MDP)

- The minimal total expected processor energy consumption in discrete time ($J^{*,\mathbb{N}}$) from 0 to T :

$$J^{*,\mathbb{N}}(w_0) = \min_{\pi} \left\{ \mathbb{E} \left(\sum_{t=0}^T Q(\pi_t(w_t)) \right) \right\} \quad (1)$$

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- Backward induction to compute $J^{*,\mathbb{N}}$:

$P_n(a)$ = occurrence probability of the work quantity arrival a .

$$J_{n-1}^{*,\mathbb{N}}(w) = \min_{s \in \mathcal{A}(w)} \left\{ Q(s) + \sum_a P_n(a) J_n^{*,\mathbb{N}}(\text{Succ}(w, s, a)) \right\} \quad (2)$$

- Admissible speed set $\mathcal{A}(w)$ to guarantee that each job is executed before its deadline:

$$\mathcal{A}(w) = \{s \in \mathcal{S} \mid s \geq w(1)\}$$

Continuous Time Control (CTC)

Discrete Time Control (DTC)

- The processor may change its speed at **integer** time $n \in \{0, 1, \dots, T\}$.
- **Integer** time evolution of the state:

$$\forall u \in \mathbb{N}, \quad w_{n+1}(u) = \text{Succ}(w_n(u), \pi(w_n), a_n(u))$$

Continuous Time Control (CTC)

- The processor may change its speed at **any** time $t \in [0, T]$.
- **Continuous** time evolution of the state:

$$(1) \quad n = 0: \forall u \geq 0,$$

$$w_0^\pi(u) = 0.$$

$$(2) \quad n > 0: \forall t \in]n, n + 1],$$

$$w_t^\pi(u) = \left(w_n^\pi(u) - \int_n^t \pi(w_v^\pi) dv \right)^+ + a_t(u).$$

CTC: Consecutive Speeds Set

Theorem

*If the set S is made of **consecutive** speeds (i.e. $S = \{0, 1, 2, \dots, s_{\max}\}$) and the power function Q is increasing and convex, then there is no energy gain for the processor to **change its speed at non-integer times**.*

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- Consider that the processor can change its speed at times $t \in \mathbb{N}$ as well as at times $t + 1/2$.

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- By continuity of the total energy:

Changing its speed at integer times is *always optimal* with a *consecutive* set of speeds.

CTC: Sketch of the Proof (i)

Goal: proving that changing speed at $t + 1/2$, with $t \in \mathbb{N}$ does not improve the energy consumption.

- Recall of the dynamic programming equation:

$$J_t^*(w) = \min_{\mathbf{s} \in \mathcal{A}(w)} \left\{ Q(\mathbf{s}) + \sum_a P_t(a) J_{t+1}^* \left(\text{Succ}(w, \mathbf{s}, a) \right) \right\} \quad (3)$$

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- Combining the two dynamic programming equations at t and $t + 1/2$:

$$\begin{aligned} J_t^*(w) = & \min_{u_{[0,1/2]} \in \mathcal{A}_1(w), u_{[1/2,1]} \in \mathcal{A}_2(w)} \left(Q \left(\frac{u_{[0,1/2]} + u_{[1/2,1]}}{2} \right) \right. \\ & \left. + \sum_a P_t(a) J_{t+1}^* \left(\text{Succ} \left(w, \frac{u_{[0,1/2]} + u_{[1/2,1]}}{2}, a \right) \right) \right) \quad (4) \end{aligned}$$

where $\mathcal{A}_2(w) = \{s \in \mathcal{S} : s \geq 2w(1)\}$, $\mathcal{A}_1(w) = \mathcal{S}$.

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- Changing speed at half times is equivalent to choosing half speeds at integer times.

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- There exists an optimal **deterministic** speed policy by MDP theory.
 \Rightarrow Here, it uses on $[t, t + 1]$ either speed u or speed $u + 1$.
- $u_{[1/2,1]} = u_{[0,1/2]} + 1$ is not strictly better than $u_{[1/2,1]} = u_{[0,1/2]}$.

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- By continuity of the total energy:

Changing speeds at **integer times** is **always optimal**
with a **consecutive** set of speeds.

Extensions

- ① Non consecutive speed set \mathcal{S} :

Emulating **non available** integer speeds with V_{dd} -hopping by using two **neighboring** $s \in \mathcal{S}$.

- ② Non convex power function $Q(\cdot)$:

Replacing the power function $Q(\cdot)$ by its **convex hull** $\hat{Q}(\cdot)$.

Conclusion

- + Minimizing in continuous time the energy consumption of a processor with a discrete set of speeds executing jobs with discrete features (arrival times, sizes and deadlines).
- + The continuous time optimization problem can be solved in discrete time.
- + In practice, the discrete optimal speed policy can be computed effectively using a finite Markov Decision Process.
 - The size of the state space of the MDP grows exponentially fast with Δ (the maximal deadline).
- + Can be implemented in an embedded system.