

Time-Constrained Energy Minimization for Online Execution of a Stochastic DAG Task

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Features of our problem

Objective

Minimize average energy consumption while ensuring that the task is completed before some hard real-time deadline.

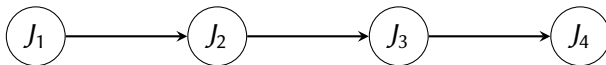
Processing Unit

- *Speed* $s \in \mathbb{R}_+$
- *Energy consumption* per unit of speed s : $Q(s) = s^{\alpha-1}$ (with $\alpha > 2$)

Deadline

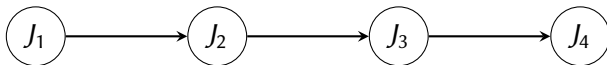
- Computation *start* at time 0
- Task **must** be completed before *deadline* $D > 0$

Task



Example of a task made of four elementary jobs

Task



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Hypothesis on jobs

- Size of job J_k in $[0, W_k^{\max}]$ with $W_k^{\max} < \infty$
- Size of jobs *unknown* until they are completed
- *Cumulative distribution* and *Tail* functions F_k and F_k^c
- *Independent*
- *Sequential* execution

Objective and existing approaches

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Our case: *Online*, and we use the *work* instead of *time*.

Problem Statement

Objective

Minimize average energy consumption while ensuring that the task is completed before the deadline.

Optimization problem

Minimize over all speed profile s the quantity $\mathcal{E}_{J_1, \dots, J_K}^s(D)$ defined by

$$\mathcal{E}_{J_1, \dots, J_K}^s(D) := \mathbb{E} \left[\int_0^{W_1 + \dots + W_K} Q(s(w)) dw \right] = \mathbb{E} \left[\int_0^{W_1 + \dots + W_K} s(w)^{\alpha-1} dw \right]$$

$$\text{subject to: } \int_0^{W_1^{\max} + \dots + W_K^{\max}} \frac{dw}{s(w)} \leq D.$$

One job

The problem can be rewritten as

$$\mathcal{E}_{J_1}^*(D) := \inf_{s_1: [0, W_1^{\max}] \rightarrow \mathbb{R}_+} \int_0^{W_1^{\max}} s_1(w)^{\alpha-1} F_1^c(w) dw.$$

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Known from the literature [LORCH and SMITH, 2001]:

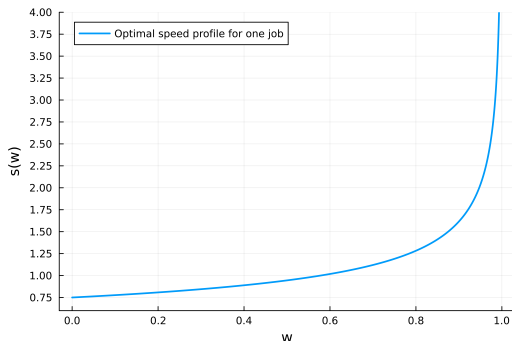
$$s_1^*(w) = \frac{C_1}{D F_1^c(w)^{\frac{1}{\alpha}}}$$

where

$$C_1 = \int_0^{W_1^{\max}} F_1^c(w)^{\frac{1}{\alpha}} dw.$$

We get

$$\mathcal{E}_{J_1}^*(D) = \frac{C_1^\alpha}{D^{\alpha-1}}.$$



Optimal speed profile for a job given by a uniform distribution on $[0, 1]$.

Several jobs is challenging

- **Adaptive** policy : The jobs are considered as one, and the speed profile is refreshed when any job is completed.
- ***d*-Sweep** policy : A deadline is fixed for each job and compute the optimal partial deadline instead.
It was used by [LU, ZHANG, STAN, LACH and SKADRON, 2006] and claimed to be optimal without proof.

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Our approach: Compute explicitly using the optimality principle

Optimality principle for two jobs

$$\mathcal{E}_{J_1, J_2}^{s_1}(D) = \int_0^{W_1^{\max}} s_1(w)^{\alpha-1} F_1^c(w) dw + \dots$$

Optimality principle for two jobs

$$\mathcal{E}_{J_1, J_2}^{s_1}(D) = \int_0^{W_1^{\max}} s_1(w)^{\alpha-1} F_1^c(w) dw + \int_0^{W_1^{\max}} \mathcal{E}_{J_2}^* \left(D - \int_0^w \frac{du}{s_1(u)} \right) dF_1(w)$$

Optimality principle for two jobs

Theorem

$$\mathcal{E}_{J_1, J_2}^*(D) = \inf_{s_1: [0, W_1^{\max}] \rightarrow \mathbb{R}_+^*} \left(\int_0^{W_1^{\max}} s_1(w)^{\alpha-1} F_1^c(w) dw + \int_0^{W_1^{\max}} \mathcal{E}_{J_2}^* \left(D - \int_0^w \frac{du}{s_1(u)} \right) dF_1(w) \right)$$

Two jobs : Optimal solution

Theorem (Optimal speed profile and mean energy consumption for two jobs)

$$s_1^*(w) = \frac{1 + \int_0^{W_1^{\max}} y(u) du}{D y(w)} \quad \text{and} \quad s_2^*(w) = \frac{C_2}{\left(D - \int_0^{w_1} \frac{du}{s_1^*(u)}\right) F_2^c(w)^{\frac{1}{\alpha}}}$$

under $[W_1 \leq w_1]$ and y is a solution of the following second order ODE (with $\int y, y, y'$):

$$\frac{y'(w)}{y(w)} = \frac{1}{\alpha} \frac{dF_1(w)}{F_1^c(w)} \left(C_2^\alpha \left(\frac{y(w)}{1 + \int_w^{W_1^{\max}} y(u) du} \right)^\alpha - 1 \right)$$

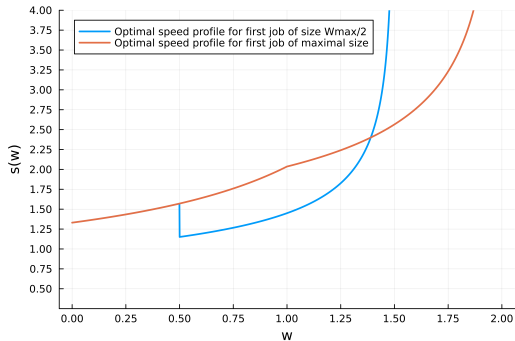
with initial condition $y(W_1^{\max}) = C_2^{-\frac{1}{\alpha}}, y'(W_1^{\max}) = 0$. In particular,

$$\mathcal{E}_{J_1, J_2}^*(D) = \frac{C_{1,2}}{D^{\alpha-1}}$$

where $C_{1,2}$ is a constant that does not depend on D .

Online example with two jobs

Job sizes W_1, W_2 follows uniform distribution on $[0, 1]$ and $D = 1$.

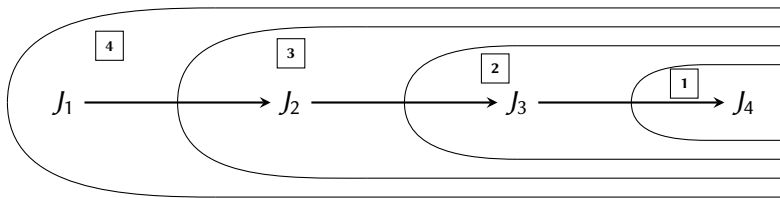


Speed profile given work executed

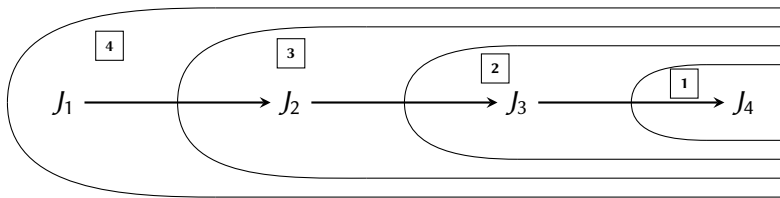


Work advancement given time

Extension to tasks of $K \geq 2$ sequential jobs



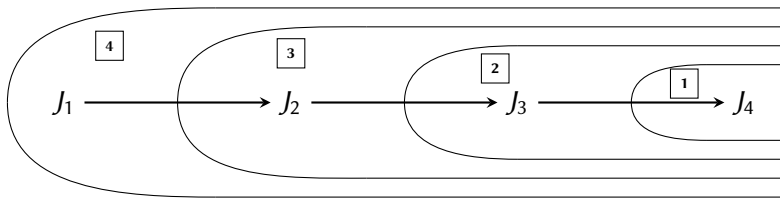
Extension to tasks of $K \geq 2$ sequential jobs



Computing $C_{k,\dots,K}$ an *offline constant by Backward Induction*

$$\begin{aligned} \mathcal{E}_{J_k,\dots,J_K}^*(D) &= \frac{C_{k,\dots,K}}{D^{\alpha-1}} \\ &= \inf_{s_k} \left\{ \int_0^{W_k^{\max}} s_k(w)^{\alpha-1} F_k^c(w) dw + \int_0^{W_k^{\max}} \frac{C_{k+1,\dots,K}}{(D - \int_0^w \frac{du}{s_k(u)})^{\alpha-1}} dF_k(w) \right\} \end{aligned}$$

Extension to tasks of $K \geq 2$ sequential jobs

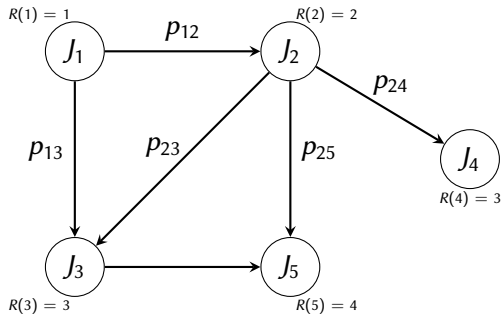


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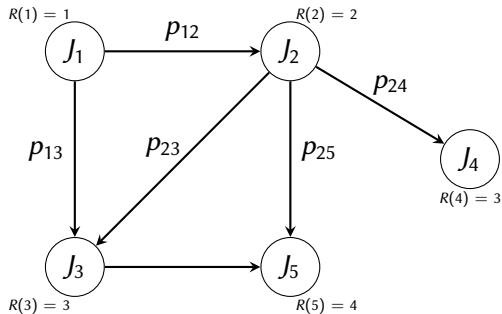
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Then computing the *online* optimal speed by *Forward induction*.

Solving flowcharts

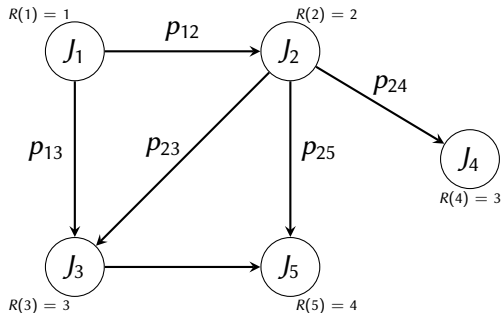


Solving flowcharts



$$\mathcal{E}_{G_1}^*(D) = \inf_{s_1} \left(\int_0^{W_1^{\max}} s_1(w)^{\alpha-1} F_1^c(w) dw + \int_0^{W_1^{\max}} \frac{\sum_{j \in \text{Succ}(1)} p_{1j}(w) C_{G_j}(W_j^{\max})^\alpha}{\left(D - \int_0^w \frac{du}{s_1(u)}\right)^{\alpha-1}} dF_1(w) \right)$$

Solving flowcharts



$$\mathcal{E}_{G_1}^*(D) = \inf_{s_1} \left(\int_0^{W_1^{\max}} s_1(w)^{\alpha-1} F_1^c(w) dw + \int_0^{W_1^{\max}} \frac{\sum_{j \in \text{Succ}(1)} p_{1j}(w) C_{G_j}(W_j^{\max})^\alpha}{\left(D - \int_0^w \frac{du}{s_1(u)}\right)^{\alpha-1}} dF_1(w) \right)$$

Global complexity is of order of $O(AN)$ where A denotes the number of arcs of the DAG and N denotes the partition size of the discretization.