Online Flow Scheduling with Deadline for Energy Conservation in Data Center Networks

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- Backgrounds
- Problem
- Solutions
- Evaluation



Backgrounds

an annual growth of 12% (U.S.Environmental Protection Agency)

Benefits:

➢ Economics:

Reduce bills for energy consumption.

Environmental Protection:

Reduce carbon emission.



Data Centers



Backgrounds



Reducing the energy consumption has becoming a key component in building green datacenters!



Backgrounds

- Q1: how to reduce energy?
- A1: speed scaling.
- Q2: how to guarantee quality of service?
- A2: ensuring flow deadline.



Challenges

- Online scheduling
 - Make decisions continuously for the arriving new flows.
- Incorporating future information
 - The predicted future information should be incorporated in order to improve the quality of the scheduling.
- Efficient in practice
 - Target the average cases rather than devoting themselves to the worst-case inputs.



Motivation example



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Data center flow scheduling

• DCN topology can be abstracted as a undirected graph.



- A flow is associated with four parameters.
 - Flow demand, release time & deadline, the source & destination, the routing path

A schedule is called "feasible" if every flow can be accomplished within its deadline following this schedule.



Models

- Energy saving mechanism
 - Speed scaling
 - $g(x_e) = \sigma + \mu x_e^{\alpha}$
- Prediction and uncertainty
 - probability-based model $w_{t,t_1} \sim \widehat{f}_{t_p,t,t_1}(w).$
 - $\widehat{\mathbf{w}}_{\beta}(t_p, t) = \{ \widehat{w}_{t_p, t, t'}^{\beta} | t' \in [t, T] \},$ $\widehat{w}_{t_p, t, t'}^{\beta} \in \{ w | \widehat{f}_{t_p, t, t_1}(w) \ge \beta \}.$





Abstracting the problem

- The Deadline-constrained Energy-efficient Flow Scheduling (DEFS) problem
 - Can we find a feasible online scheduling such that the total energy consumption is minimized?

$$\Phi_f(\mathcal{S}) = T \cdot |\mathcal{E}_a| \cdot \sigma + \int_0^T \sum_{e \in \mathcal{E}_a} \mu(x_e(t))^\alpha dt.$$

• Solving the problem is NP-hard!



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The computation of scaling factor

Lemma 1. ([24]) An online algorithm, A, is feasible iff for all $Z \in \mathcal{Z}$ and all $t_1 \leq t_2$, $t_1, t_2 \in \mathcal{T}$, the following inequality holds:

$$\sum_{t=t_1}^{t_2} \sum_{s=t}^{t_2} w_{t,s} \le \int_{t_1}^{t_2} x_e(t) dt.$$
(8)

Combining $x_e(t) \leq \sigma_{t_1,t_2} x_e^*(t)$ with Lemma 1,

$$\sum_{t=t_1}^{t_2} \sum_{s=t}^{t_2} w_{t,s} \le \int_{t_1}^{t_2} x_e(t) dt \le \sigma_{t_1,t_2} \int_{t_1}^{t_2} x_e^*(t) dt$$



The computation of scaling factor

According to prediction information and by applying Yao's algorithm, we have

$$x_e^*(Z_t^\beta) = \inf_{Z' \in \mathcal{Z}_\beta, Z'_t = Z_t^\beta} x'_e(Z_t^\beta).$$

Substituting $x_e^*(t)$ with $x_e^*(Z_t^\beta) \le x_e^*(t)$ for inequation

$$\sigma_{t_1,t_2} \ge \frac{\sum_{t=t_1}^{t_2} \sum_{s=t}^{t_2} w_{t,s}}{\int_{t_1}^{t_2} x_e^*(Z_t^\beta)}.$$

By solving the following optimization problem,

$$\sigma_{t_1,t_2} = \sup_{Z \in \mathcal{Z}_\beta} \frac{\sum_{t=t_1}^{t_2} \sum_{s=t}^{t_2} w_{t,s}}{\int_{t_1}^{t_2} x_e^*(Z_t^\beta)}$$

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The evaluation setting

- Data center topology
 - A 8-ary Fat-Tree with 128 servers
- Benchmarks
 - Our algorithm (TRO)
 - Most-Critical-First (MCF)
 - Online scheduling without prediction (BKP)
- Evaluation metric
 - Total energy consumption.



Single VC request scenario

Case	MCF	DEFSA-I	DEFSA-II	DEFSA-III	BKP
100#	3	91	63	90	336
350#	10171	405494	205976	325494	1086340

Case	DEFSA-L0	DEFSA-L30	DEFSA-L60
100#	282	63	23
350#	995307	205976	91342



Multiple VC requests scenario



Thanks for your attention!

