

1 Optimal scheduling for m homogeneous servers

TODO: introduction text

1.1 Input and conventions

TODO: rewrite in nice paragraph

Input:

- $m \in \mathbb{N}$: Number of homogeneous servers
- $T \in \mathbb{N}$: Number of time steps
- $\beta \in \mathbb{R}_{\geq 0}$: Power up costs
- $\lambda_0, \dots, \lambda_T \in [0, m]$: Arrival rates

Notations:

- Let $\lambda_{i,t}$ be the assigned arrival rate at time t for server i
- Let x_t be the number of active servers at time t
- Let $\mathcal{X} := (x_0, \dots, x_T)$ be the sequence of active servers

Requirements:

- Convex cost function f
- Power down costs are w.l.o.g. equal to 0
- $\forall t \in \{0, \dots, T\} : \sum_{i=1}^m \lambda_{i,t} = \lambda_t$
- $\lambda_0 = \lambda_T = 0$
- $\mathcal{X}(0) = \mathcal{X}(T) = 0$, i.e. all servers are powered down at $t = 0$ and $t = T$

1.2 Preliminaries

Lemma 1.1. *Given a convex cost function f , x active servers and an arrival rate λ , the best method is to assign each server a load of λ/x .*

Proof. $\forall x \in \mathbb{N}, \mu_i \in [0, 1] : \sum_{i=1}^x \mu_i = 1 :$

$$\begin{aligned} f\left(\frac{\lambda}{x}\right) &= f\left(\sum_{i=1}^x \frac{\mu_i * \lambda}{x}\right) \stackrel{\text{Jensen's inequality}}{\leq} \sum_{i=1}^x \frac{1}{x} f(\mu_i * \lambda) \\ &\Leftrightarrow x * f\left(\frac{\lambda}{x}\right) \leq \sum_{i=1}^x f(\mu_i * \lambda) \end{aligned}$$

□

We will use this fact in our following construction for an optimal schedule.

1.3 Graph for an optimal schedule

We construct a directed acyclic graph as follows:

$\forall t \in [T - 1]$ and $i, j \in \{0, \dots, m\}$ we add vertices (t, i) modelling the number of active servers at time t . Furthermore, we add vertices $(0, 0)$ and $(T, 0)$ for our initial and final state respectively. In order to warrant that there are at least $\lceil \lambda_t \rceil$ active servers $\forall t \in [T - 1]$, we define an auxiliary function which calculates the costs for handling an arrival rate λ with x active servers:

$$c(x, \lambda) := \begin{cases} x * f(\lambda/x), & \text{if } \lambda \leq x \\ \infty, & \text{otherwise} \end{cases} \quad (1)$$

Then, $\forall t \in [T - 2]$ and $i, j \in \{0, \dots, m\}$ we add edges from (t, i) to $(t + 1, j)$ with weight

$$d(i, j, \lambda_{t+1}) := \underbrace{\beta * \min\{0, j - i\}}_{\text{power up costs}} + c(j, \lambda_{t+1}) \quad (2)$$

Finally, for $0 \leq i \leq m$ we add edges from $(0, 0)$ to $(1, i)$ with weight $d(0, i, \lambda_1)$ and from $(T - 1, i)$ to $(T, 0)$ with weight $d(i, 0, \lambda_T) = 0$.

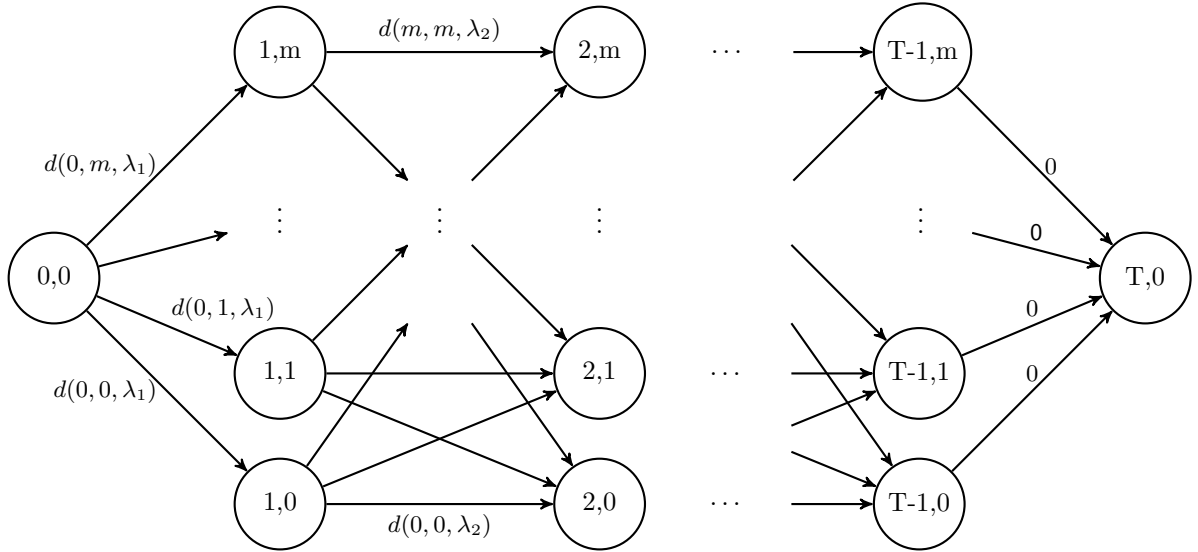


Figure 1: Graph for optimal schedule algorithm.

Note: All edges from (t, i) to $(t + 1, j)$ have weight $d(i, j, \lambda_{t+1})$

1.4 Proof of correctness

Proposition 1.2. *Any given optimal schedule corresponds to a shortest path from $(0, 0)$ to $(T, 0)$ in the constructed graph and vice versa.*

Proof.

“ \Rightarrow ”: Lemma 1.1 shows that in an optimal schedule each arrival rate λ_t will be shared equally on each active server at time t . Therefore, we can denote an optimal schedule uniquely by the sequence \mathcal{X} of active servers.

We can construct a valid path in our graph from \mathcal{X} as follows:

$\forall t \in [T]$ set $e_t := ((t-1, \mathcal{X}(t-1)), (t, \mathcal{X}(t)))$. Then set $P := (e_1, \dots, e_T)$.

As each edge e_t in our graph has weight $d(\mathcal{X}(t-1), \mathcal{X}(t), \lambda_t)$ and hence corresponds to the costs of switching from $\mathcal{X}(t-1)$ to $\mathcal{X}(t)$ servers and processing λ_t with $\mathcal{X}(t)$ active servers, it directly follows that P is a shortest path of the graph.

“ \Leftarrow ”: Let $P = ((0, 0) = v_0, \dots, v_T = (T, 0))$ with $v_t \in \{(t, i) \mid 0 \leq i \leq m\}$ be a shortest path of the graph. Again, it follows from lemma 1.1 that an optimal schedule is uniquely identified by the sequence \mathcal{X} of active servers.

We can construct a schedule from P by setting $\mathcal{X} = (v_0(1), \dots, v_T(1))$ where $v_t(1)$ is the second component of the t -th tuple in P .

By definition (1) it is guaranteed that P only traverses edges such that there are enough active servers $\forall t \in [T]$. Therefore, the created schedule is feasible. It's optimality directly follows from the definition of the edges' weights.

□

1.5 A minimum cost algorithm

Algorithm 1 Calculate costs for m homogeneous servers

Require: Convex cost function f , $\lambda_0 = \lambda_T = 0$, $\forall t \in [T-1] : \lambda_t \in [0, m]$

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1: function SCHEDULE( $m, T, \beta, \lambda_1, \dots, \lambda_{T-1}$ )
2:   if  $T < 2$  then return
3:   let  $p[2 \dots T-1, m]$  and  $M[1 \dots T-1, m]$  be new arrays
4:   for  $j \leftarrow 0$  to  $m$  do
5:      $M[1, j] \leftarrow d(0, j, \lambda_1)$ 
6:   for  $t \leftarrow 1$  to  $T-2$  do
7:     for  $j \leftarrow 0$  to  $m$  do
8:        $opt \leftarrow \infty$ 
9:       for  $i \leftarrow 0$  to  $m$  do
10:         $M[t+1, j] \leftarrow M[t, i] + d(i, j, \lambda_{t+1})$ 
11:        if  $M[t+1, j] < opt$  then
12:           $opt \leftarrow M[t+1, j]$ 
13:           $p[t+1, j] \leftarrow i$ 
14:         $M[t+1, j] \leftarrow opt$ 
15:   return  $p$  and  $M$ 

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Algorithm 2 Extract schedule for n homogeneous servers

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1: function EXTRACT( $m, p, M, T$ )
2:   let  $x[0 \dots T]$  be a new array
3:    $x[0] \leftarrow x[T] \leftarrow 0$ 
4:   if  $T < 2$  then return  $x$  ▷ Trivial solution
5:    $x[T-1] \leftarrow \arg \min_{0 \leq i \leq m} \{M[T-1, i]\}$ 
6:   for  $t \leftarrow T-2$  to  $1$  do
7:      $x[t] \leftarrow p[t+1, x[t+1]]$ 
8:   return  $x$ 

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1.5.1 Runtime analysis

Schedule: Loop 5,8 and 10 run $m + 1$ times, loop 7 runs $T - 2$ times

Extract: Loop 5 runs $T - 2$ times, argmin 4 takes time $m + 1$.

For $T, n \rightarrow \infty$ it holds:

$$\mathcal{O}(m + 1 + (T - 2) * (m + 1)^2 + T - 2 + m + 1) = \mathcal{O}(2 * m + T + (T - 2) * (m + 1)^2) = \mathcal{O}(T * m^2) \quad (3)$$

As we need $\log_2(m)$ bits to encode m , the algorithm is exponential in the number of servers.

1.6 A memory optimized algorithm

TODO

2 A 3/2-approximation algorithm for monotonically increasing convex f

We consider a modification of the problem discussed in chapter 1. Assuming that f is convex and monotonically increasing, we can modify our algorithm to obtain a polynomial time 3/2-approximation algorithm.

2.1 Graph for a 3/2-optimal schedule

The idea of the construction is to reduce the number of vertices by stop adding m vertices for each timestep but using vertices that approximate the number of active servers.

First, let $b := \lceil \log_2(m) \rceil$. We add vertices $(t, 0)$ and $(t, 2^i), \forall t \in [T - 1], i \in \{0, \dots, \lceil \log_2(m) \rceil\}$.

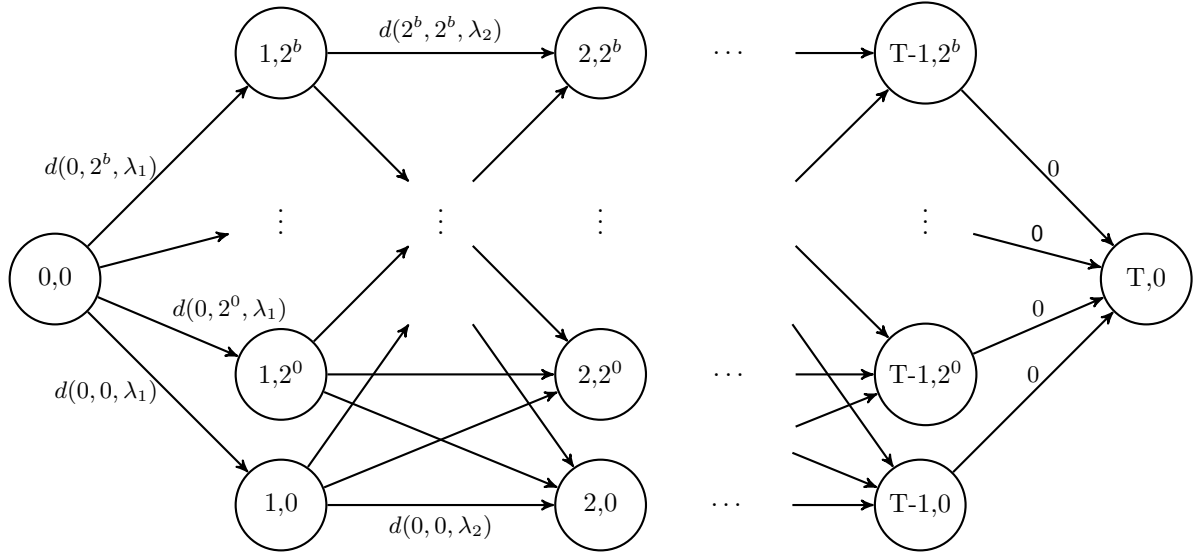


Figure 2: Graph for 3/2-approximation algorithm

Proposition 2.1. *A shortest path from $(0, 0)$ to $(T, 0)$ in the constructed graph delivers a 3/2-optimal schedule.*

Proof. Assume we have an optimal schedule identified by $\mathcal{X} = (x_0, \dots, x_T)$. □