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
- $\lambda_0, \ldots, \lambda_T \in [0, m]$: 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:

- \bullet Convex cost function f
- Power down costs are w.l.o.g. equal to 0
- $\forall t \in \{0,\ldots,T\} : \sum_{i=1}^{m} \lambda_{i,t} = \lambda_t$
- $\lambda_0 = \lambda_T = 0$
- $\mathcal{X}(0) = \mathcal{X}(T) = 0$, 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. $\forall x \in \mathbb{N}, \mu_i \in [0, 1] : \sum_{i=1}^{x} \mu_i = 1 :$

$$f\left(\frac{\lambda}{x}\right) = f\left(\sum_{i=1}^{x} \frac{\mu_i * \lambda}{x}\right) \xrightarrow{\text{Jensen's inequality}} \sum_{i=1}^{x} \frac{1}{x} f(\mu_i * \lambda)$$

$$\Leftrightarrow x * f\left(\frac{\lambda}{x}\right) \leq \sum_{i=1}^{x} f(\mu_i * \lambda)$$

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 $i, j \in \{0, ..., m\}$ 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:

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

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

Finally, for $0 \le i \le m$ 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 $d(i,0,\lambda_T) = 0$.

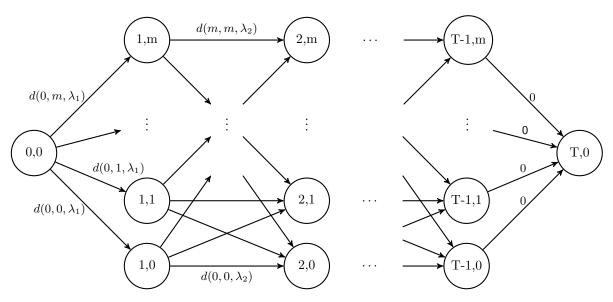


Figure 1: All edges from (t, i) to (t + 1, j) have weight $d(i, j, \lambda_{t+1})$

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:

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

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 $P = ((0,0) = v_0, \ldots, v_T = (T,0))$ with $v_t \in \{(t,i) \mid 0 \le i \le m\}$ be a shortest path of the graph. Again, it follows from equation (??) 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]
 1: function SCHEDULE(m, T, \beta, \lambda_1, \dots, \lambda_{T-1})
         if T < 2 then return
 2:
         let p[2...T-1,m] and M[1...T-1,m] be new arrays
 3:
         for j \leftarrow 0 to m do
 4:
             M[1,j] \leftarrow d(0,j,\lambda_1)
 5:
         for t \leftarrow 1 to T-2 do
 6:
             for j \leftarrow 0 to m do
 7:
                  opt \leftarrow \infty
 8:
                  for i \leftarrow 0 to m do
 9:
                      M[t+1,j] \leftarrow M[t,i] + d(i,j,\lambda_{t+1})
10:
                      if M[t+1,j] < opt then
11:
                          opt \leftarrow M[t+1,j]
12:
                          p[t+1,j] \leftarrow i
13:
                  M[t+1,j] \leftarrow opt
14:
         return p and M
15:
```

Algorithm 2 Extract schedule for n homogeneous servers

```
1: function Extract(m, p, M, T)
        let x[0...T] be a new array
2:
3:
        x[0] \leftarrow x[T] \leftarrow 0
        if T < 2 then return x
                                               ▶ Trivial solution
4:
        x[T-1] \leftarrow arg \ min\{M[T-1,i]\}
5:
                        0 \le i \le m
        for t \leftarrow T - 2 \, \bar{\text{to}} \, 1 \, \, \mathbf{do}
6:
            x[t] \leftarrow p[t+1, x[t+1]]
7:
        return x
8:
```

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 $T, n \to \infty$ it holds:

$$\mathcal{O}(m+1+(T-2)*(m+1)^2+T-2+m+1) = \mathcal{O}(2*m+T+(T-2)*(m+1)^2) = \mathcal{O}(T*m^2)$$
(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 minimize the layers of the graph by using vertices