# Split proof scheduling

### 1 Setting

There are n jobs with sizes  $p \in \mathbb{R}^n_+$  to be processed on a server with capacity 1. Jobs arrive at times  $d_i$  (unknown to the server) and are completed by some time  $C_i$ , and we are interested in minimizing expected total flow

$$F = \sum F_i = \sum (C_i - d_i).$$

We say scheduling mechanism is split proof if max waiting time of any two jobs is at least the waiting time of a single job with size equal two the sum of sizes of the two jobs.

### 2 Exponential Scheduling

Consider the following online scheduling algorithm. When a job of size  $p_i$  arrives, we generate a random score  $v_i \sim exp(\lambda = p_i)$ . Then at any time the server processes the job with highest score.

Additionally, scores of jobs that were processed for some time t' are adjusted as  $v'_i = \frac{vp_i}{p_i - t'}$ . Computationally, we only need to do this reassignment when a new job arrives. Also note that for any t' the distribution for  $v'_i$  is  $v'_i \sim exp(\lambda = p_i - t')$  (scaling property of exponential distribution).

**Theorem 2.1.** Exponential scheduling is split-proof

*Proof.* This is a pretty simple argument

## 3 Flow guarantees for exponential scheduling

**Theorem 3.1.** When  $d_i = 0$  for all i, exponential scheduling gives a 2-approximation for optimal flow.

*Proof.* We prove this by proving 2-approximation for total delay,  $D = \sum (C_i - p_i)$ .

$$D = \sum_{i} \sum_{j \neq i} D_{ij},$$

where  $D_{ij}$  is delay that job j imposes on job i. Suppose jobs are sorted in ascending order. Than, in optimal schedule we have  $D_{ij} + D_{ji} = min(p_i, p_j)$ .

For exponential scheduling, for any two jobs, we have  $\mathbb{P}[i>j] = \frac{p_i}{p_i + p_j}$ . (i>j means i is scheduled after j). Thus, we have  $D_{ij} + D_{ji} = \frac{2p_i p_j}{p_i + p_j} \leq 2min(p_i, p_j)$ . Done.

Unfortunately, this does not extend to arbitrary release dates. Consider the following instance: on t = 1, n jobs are released, n - 1 of them are of size 1 and the last one is of size A. Then, on turn t = n - 1 new jobs of size 1 start arriving at every integer time, for a long-long time. For large enough n, w.h.p. exponential scheduling would result in a queue of size A by turn t = n - 1 and so there is no constant factor approximation.

Conjecture 3.2. If server is sped up by a factor of 2, exponential scheduling is 1-competitive with the optimal schedule.

(Really, I would be happy with any constant factor approximation and any speed up).

#### 3.1 Some useful facts

I list a few things that I tried in case they're useful.

- 1) One could try to show, that for significantly large speed-up,  $N_t$ , number of jobs in queue at times t is always smaller for the exp scheduling. This is unfortunately, not true, the instance that breaks it for any speed-up is  $p_i = i$ .
- 2) Since we know that Prus's algo that processes the least processed job is competitive under speed-up, we can try and show that exp scheduling is competitive with it rather than with SRPT. I couldn't find a way to use this.
- 3) It turns out that a simpler randomized algo is provably always worse than exp scheduling and might be simpler to analyze: for every job you sample it's virtual size  $p_i' \sim u[0, p_i]$  and process shortest remaining virtual size. Couldn't do anything with this either.