

# File Delivery with Longest Processing Time First Scheduling in P2P Networks

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## Abstract

*Peer-to-peer (P2P) File delivery systems of distributed computing are designed with the understanding that any peer can leave the network at any time, often right after completing its download. The conventional approach to P2P scheduling, shortest processing-time first, is not well suited to peer-leaving situations. We therefore propose a new scheduling algorithm that significantly reduces average finish time in such cases. We further note, at the end of the paper, that our method is also suitable for the more general problem of a dynamic network of peers that may leave early or enter late.*

**Keywords:** P2P, scheduling, data sharing

## 1. Introduction

Most P2P schedulers (e.g., [1, 2, 3, 4, 5, 6]) prefer those peers with shortest processing times (i.e., higher upload bandwidths). This is called SPT scheduling [7, 8]. But peers often disconnect once their downloads finish, to save power or bandwidth [9]. If this happens then SPT scheduling performs poorly, with the total bandwidth of upload ( $BW_k$ ) rapidly dropping off in each time interval. It is instead better to do the opposite, by favoring peers with longest processing times (LPT scheduling [5]). But this also is non-ideal: since fast peers receive data slowly, their contribution  $BW_k$  is limited.

Alternatively, we might use SPT scheduling but withhold the fast peers in the network to service others. But that would be unwise, because all peers would finish together, and increase the average finish time,  $T_A$ .  $T_A$  is an important metric of overall performance, but minimizing it is considered NP-complete [10, 11, 12].

We instead propose a novel scheduler, Reverse Store-and-Forward (RSF), which unicasts data and uses inverse-multiplexing [13] with divide-and-conquer approach. By increasing the time at which the last peer finishes ( $T_L$ ), RSF substantially decreases the  $T_A$  of networks with peer-leaving.

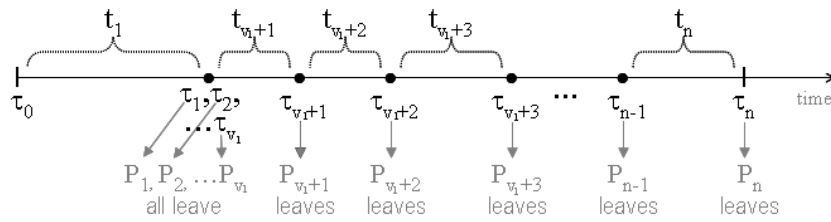
## 2. System Model

A single server ( $P_0$ ) with an upload bandwidth (or capacity) of  $C_0$  services  $n$  peers ( $P_j, j=1, \dots, n$ ) with capacities  $C_j$ . Since the labeling of the peers is arbitrary, we may choose an ordering such that the slowest is  $P_1$  and the fastest is  $P_n$  (i.e.,  $C_1 \leq C_2 \leq \dots \leq C_n$ ), such that  $BW_k$  is maximized when peers are leaving sequentially from  $P_1$  to  $P_n$ . Furthermore, at time  $\tau_0$ , each of the  $n$  peers requests a data file (of size  $|F|$ ) from  $P_0$ . We adopt the assumptions of [1]: no new peer arrivals, no delays in data redistribution, no repeated data transmissions, a fully connected network, and no constraints on download (retrieving) capacity. We also add the

assumption that each peer leaves immediately after it obtains a complete copy of the file. A proof for our more-difficult problem of peer-leaving is harder, however. Still, given that files sent in P2P networks tend to be large, this assumption often holds.

As Figure 1 shows, RSF scheduling creates time intervals, each ending when some peer leaves. Figure 1 also labels the time intervals. During the  $k_{th}$  such interval, peer  $P_k$  is at the head of the scheduling queue. Then, at time  $\tau_k$ , this peer  $P_k$  finishes. Only at these discrete times does  $BW_k$  reduce, being constant within each interval. The higher  $BW_k$  gives short time interval.

Moreover, since peers are ordered by capacity, Figure 1 also shows that slower peers finish first: after  $k-1$  peers have left, only those peers of higher number remain ( $P_k, P_{k+1}, \dots, P_n$ ). In this way, the finishing time ( $\tau_k$ ) of each  $P_k$  is, in turn (and in order of slowness), greedily minimized without considering its effect on future intervals.



**Figure 1. The timeline of peer-leaving events in the proposed RSF Scheduling**

Such notation is useful because we apply the Min-Min heuristic of [1], which simplifies the problem of minimizing  $T_A$  across all peers to the smaller problem of minimizing only  $t_k$  ( $= \tau_{k+1} - \tau_k$ ). Note that some  $t_k$  can be zero (e.g.,  $t_2$  or  $t_{v_1}$  of Figure 1). Once  $P_k$  finishes, the goal then changes to minimizing  $t_{k+1}$ . Then, the relationship between  $T_A$  and  $t_k$  is:

$$T_A = \frac{\tau_1 + \tau_2 + \dots + \tau_n}{n} = \frac{\sum_{k=1}^n (n-k+1)t_k}{n} \quad (1)$$

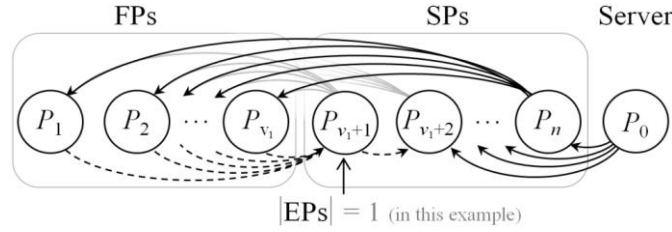
We also use the following notations:

- $C_0$ : the capacity of the source provider
- $C_i$ : the capacity of peer  $P_i$
- $BW_k$ : the total capacity ( $C_0 + C_k + C_{k+1} + \dots + C_n$ ) of P2P network in the time interval of  $t_k$ . This constrains the performance of P2P file delivery.
- $r_{ij}(t_k)$ : the transmission rate from  $P_i$  to  $P_j$  during interval  $t_k$ . (From which it follows that  $\sum_{i=k}^n r_{ij}(t_k) \leq C_i \mid i \neq j$ .)
- $R_j(t_k)$ : the total bandwidth which  $P_j$  receives during interval  $t_k$ . Since the total receiving capacity cannot exceed the total upload capacity, it follows that  $R_j(t_k) \leq BW_k$ . For an example,  $\sum_{j=k}^n R_j(t_k) = r_{0j}(t_k) + \sum_{i=k}^n r_{ij}(t_k)$ .

A single-server system can be either slow ( $C_0 \leq \sum_{j=1}^n \frac{c_j}{n-1}$ ) or fast ( $C_0 > \sum_{j=1}^n \frac{c_j}{n-1}$ ) [14]. It is known that a slow-server condition is trivial to schedule, because all peers can receive data at the rate  $C_0$  [15]. We therefore only consider the fast-server condition.

### 3. Reverse Store-and-Forward Scheduling

As noted above, the RSF scheduler seeks to minimize  $t_k$ , for each value of  $k$ . This goal can be equivalently restated as minimizing the time needed for the slowest remaining peer ( $P_k$ ) to finish. One way to obtain this goal would be, of course, to have  $P_0$  only send to  $P_k$ . The problem here is that the P2P network would be significantly underutilized;  $P_{k+1}$  could receive forwarded data from  $P_k$  but only at the rate of  $C_k$  (i.e., the smallest capacity). This is the pitfall of all greedy algorithms – that minimizing the current step can make later steps take longer.



**Figure 2. Communication pathways allowed during the first time interval**

We overcome this pitfall by hybridizing divide-and-conquer method. For RSF scheduling, the configuration obtains the same minimum value for  $t_k$ , but with better network usage. RSF divides peers into two partitions: those which are receiving data from  $P_0$  and those which are receiving forwarded data from other peers. There are therefore two constraints in RSF scheduling: the just-mentioned partition-restriction, and the aforementioned requirements to greedily minimize  $t_k$ . Subject to these constraints, RSF scheduling optimizes the total P2P network bandwidth within each time interval. In other words:

$$\text{minimize: } t_k = \frac{\text{data missing from } P_k \text{ at } \tau_{k-1}}{\text{transmission rate to } P_k \text{ on } t_k} = \frac{|F| \sum_{i=1}^{k-1} t_i \times R_k(t_i)}{R_k(t_k)}$$

$$\text{subject to: } t_1 \geq \frac{F}{C_0} \quad \text{and} \quad \sum_{j=k}^n R_j(t_k) \leq BW_k$$

$$\text{given that: } R_k(t_k) = \begin{cases} C_0 & k = 1 \\ \sum_{i=k+v_k}^n \frac{C_j}{v_k} + \sum_{i=k+1}^{k+v_k-1} C_i & k > 1 \end{cases}$$

where  $v_k$  is the partition boundary:  $(P_k \dots P_{k+v_k-1}), (P_{k+v_k} \dots P_n)$

The above equations raise questions about the partition scheme. One question regards how the partition boundary,  $v_k$ , is chosen. Another question regards the purpose that these partitions serve. To simplify these questions, we will now begin by considering the pathways of the first time interval.

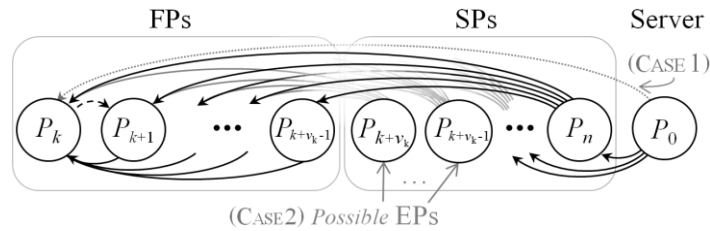
**ON THE FIRST TIME INTERVAL**,  $k=1$ , so the partitions are  $(P_1, P_2, \dots P_{v_1})$  and  $(P_{v_1+1}, P_{v_1+2}, \dots P_n)$ , as Figure 2 shows. This figure labels these partitions as FPs (for finishing peers) and SPs (for supporting peers), respectively. Additionally, there is a possibility of having EPs (for excess peers), shown as a subset of the SPs.

The distinctions between these sets are in the allowed communication pathways. From Figure 2, the primary pathway is right-to-left, as shown by the solid lines in the figure:  $P_0$  sends data to the SPs, while these SPs send data to the FPs. But these two transmissions

differ: whereas  $P_0$  unicasts data and uses IMUX to send *unique* data to each SP (except for the EPs subset), yet an SP sends the *same* data to each FP.

The dashed lines in Figure 2 show left-to-right pathways. We do not wish to waste the upload bandwidth of the FPs (although it is small). Since the FPs still have the upload bandwidth, they instead forward the received to the slowest SP.

Let us now make a brief aside to consider the value of  $\tau_1$  for the above configuration. First, consider any one peer  $P_j$ , where  $P_j \in \text{SPs}$  and  $P_j \notin \text{EPs}$ . Note that, since  $P_j$  multicasts to each FP, Since  $P_j$  cannot transmit any faster than  $C_j/v_1$ , it does not need to receive any faster, either. Second, suppose we choose the SP set to be large enough so that  $\sum_{i \in \text{SPs}} \frac{C_i}{v_1} \geq C_0$ . It then follows that  $P_0$  transmits unique data as fast as it possibly can (since  $P_0$  sends unique data to each SP). Consequently, the capacity of the server is multiplied by  $v_1$  and each FP receives unique data at a rate of  $C_0$ . As intended, this configuration is not only greedy for the finish time of  $P_1$  ( $\tau_1$  is at its theoretical minimum), but it is also effective at speeding up other peers<sup>3</sup>.



**Figure 3. Communication pathways allowed on later time intervals**

Now consider the choice of  $v_k$ , on each time interval  $t_k$ . The optimum value of  $v_k$  should minimize any mismatch between the number of FPs and the available capacity among the SPs:

$$v_k = v^*, \text{ for that } v^* \in (1 \dots n-k) \text{ that minimizes } \left| \sum_{i=v^*+k}^n C_i - v^* \times C_0 \right| \quad (2)$$

This equation has an absolute value because the difference may be negative. If negative (which we call **CASE 1**), it means that the SP and FP sets overlap:  $P_0$  has extra capacity (to give directly to  $P_k$ ). If positive (which we call **CASE 2**), then the SPs have excess capacity – meaning that the EPs set will not be empty. (The third case – that of a perfect match – is considered as a trivial form of **CASE 2**; nothing more will be said of it.) The first interval is special because **CASE 1** is disallowed.

Note that the number of EPs may not be an integer. In this case, EPs contribute their capacities to servicing the SPs. Common sense dictates the way, we hereafter simplify the discussion (without loss of generality) to assume that  $|\text{EPs}|$  (the number of EPs) is an integer.

**ON LATER TIME INTERVALS**, scheduling is more complex, as Figure 3 shows. These lines indicate that there are no EPs and that  $P_0$ 's extra bandwidth goes to  $P_k$ . This is because all of the FPs finish at the same time as  $P_1$ . Each FP still receives from the SPs. But all capacities of the FPs (except for redundant data) go to  $P_k$ . With  $\text{BW}_k$  fully utilized,  $t_k$  is minimized and the requirement of greediness is met. There are now two cases. In **CASE 1**, the additional green dotted lines are allowed. As for the dashed path in the Figure,  $P_k$ 's capacity goes to  $P_{k+1}$ . The Figure also has additional (dot line) path that are only valid in **CASE 1**.

Most importantly, these pathways are heuristic that simplifies the problem to find  $v_k$ . Equation (2) finds an approximate answer, for which the optimal answer of the specific pathways for each peer would be prohibitively difficult to make an algorithm faster, especially when the number of peers is large. When concerning a single time interval,  $t_k$  is minimized optimally, because the pathways allow  $P_k$  to receive data as fast as possible, and the problem of resource relocation is solved with a divide-and-conquer approach.

**PEER REMAINING SUPPORTED APPROACHES**, these are enhanced versions of RSF for which peers can register themselves as the designated remaining peers (RPs). In this case, peers may not leave immediately after finish receiving data. Even if most peers do leave, those that stay can have a large effect on  $T_A$ . We model the likelihood of whether a peer does remain as the  $PR_m$  (*i.e.*,  $PR_m = |RPs| / n$ ), and server has no need to send any data to RPs, the capacity of RPs do not count in (2). Additionally, we assume RPs remain till all peers finish.

Since RPs will remain, our first strategy, eRSF, is to place the RPs at the front of the queue. Thus, if  $P_m$  is a RP, then we number FPs such that  $C_m, C_k, C_{k+1}, \dots, C_{k+v_k}$ . After RPs finish receiving data, they are marked as EPs`. EPs` are part of EPs to serve FPs, but they will not turn into FPs. A user of eRSF is rewarded for remaining in the network, by getting less waiting time.

Additionally, uRSF is introduced for the incentive of large capacity. To reward a user for investing in larger capacity, we now modify eRSF by numbering all of FPs with SPT scheduling, thus  $C_k \geq C_{k+1} \geq \dots \geq C_{k+v_k}$ . Then, faster RPs (*i.e.*, RPs with larger  $C_j$ ) would finish first by place them at the front of the queue.

## 4. Results

We compare our method against two others. The first is SPT, the scheduling algorithm of [1]. SPT perfectly minimizes  $T_A$ , if peers do not leave. The second is Even-Share Scheduling, ESS, in which every peer shares equally with all others, so that they all finish together. ESS perfectly minimizes  $T_L$ , the time for the last peer to finish. This minimum  $T_L$  is constrained by the physical requirement that  $T_L^* \geq \max(|F|/C_0, n|F|/\sum_{j \in \{0,1,\dots,n\}} C_j)$ .

Figure 4 shows that RSF performs well if peers leave, and acceptably if peers remain (given that SPT is, in this case, optimal). The results are for both peer-leaving and peer-staying scenarios. For comparison, the network is the same as in [1]:  $n = 4$ ,  $BW_k = 25$ , and  $|F| = 144$ . The peer capacities are  $C_0=12$ ,  $C_1=6$ ,  $C_2=4$ ,  $C_3=2$ ,  $C_4=1$ . Also notice, there is no difference between the finish times of ESS, because all peers finish together.

We next compare the effect of peers having various distributions of capacities, with the same  $BW_k=25$  and  $C_0=12$  that were used in Figure 4. There are three basic types of distributions. The peers of RSF-extreme has capacities that differ greatly:  $C_1=1$ ,  $C_2=1$ ,  $C_3=1$ ,  $C_4=10$ . On the contrary, the peers of RSF-similar have capacities that distribute lesser:  $C_1=3$ ,  $C_2=3$ ,  $C_3=3$ ,  $C_4=4$ . The RSF-reference is reproduced from the RSF (peer-leaving) in Figure 4. In Figure 5, the RSF has the worst performance when the differences between peers' capacities is the small. Our simulations will base on RSF-similar condition to avoid over-optimistic assessment.

Subsequently, we evaluate the effect of some peers staying upon the performance of the various scheduling strategies, at varying peer sizes. The experiments assume that peer are selfish, and all  $C_i$  have an asymptotically chi-square distribution with 1 degrees of freedom, such that we always have fast-server condition and more than 90% of  $C_i \approx 1$  for lesser distributions of capacities. In addition, this experiment models the situation where peers do not always leave: by assuming that 80% of the peers are randomly selected to leave after finishing. Then, the experiment is run 1000 times. With confidence interval is 90%, we refer

the mean of the results by solid lines, and the standard deviation ( $\sigma$ ) by different types of dotted lines. Note, the result for SPT was also intended, but was too poor to graph into the same plot area, for  $n > 10$ .

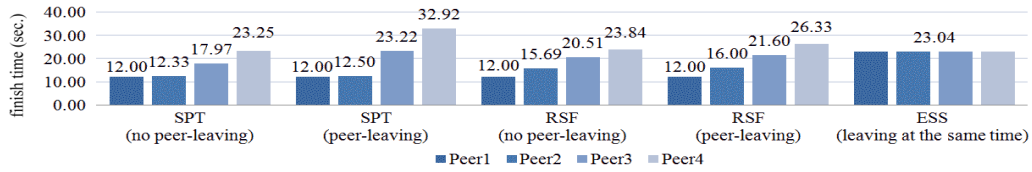


Figure 4. Comparison between RSF, SPT, and ESS

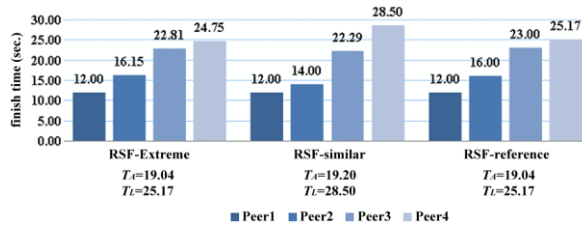


Figure 5. Comparison between RSF, SPT, and ESS.

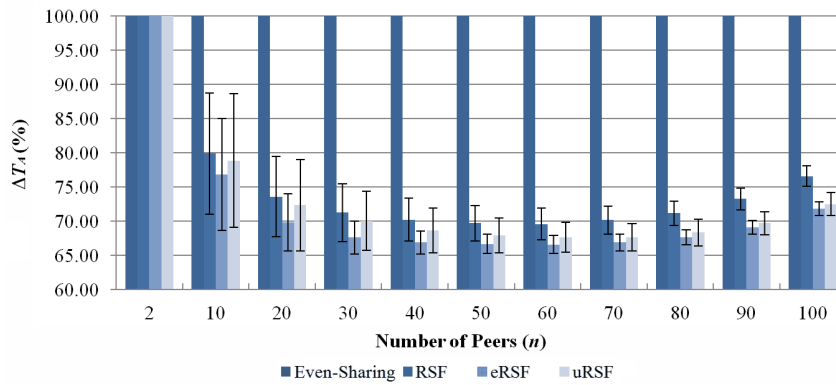


Figure 6. Performance of Normalized  $T_A$  vs. Number of Peers

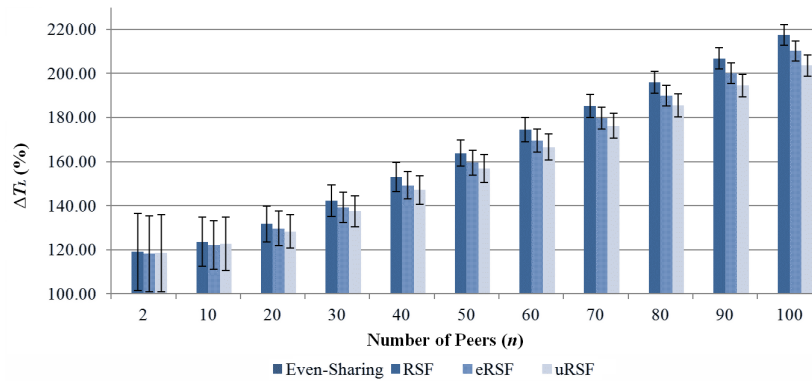


Figure 7. Performance of Normalized  $T_L$  vs. Number of Peers

In Figure 6,  $T_A$  is normalized to ESS as  $\Delta T_A$ , and  $\Delta T_A = (T_A - T_L^*) / T_L^*$ . Not surprisingly, the figure shows that ESS is an upper bound for  $\Delta T_A$ . RSF, as expected, performs worse (*i.e.*, smaller  $\Delta T_A$  is better) than eRSF and uRSF. This is because RSF cannot benefit from peer-remaining. Also, the unexpected, uRSF (which is the only method to consider SPT scheduling) has worse result of  $\Delta T_A$  than eRSF. This happens because removing some fast peers from the tail of the SPs queue will weaken the bandwidth of getting the data from the server and  $\tau_1$  is irreparably belated. Although this is unfortunate, recall that the purpose of the uRSF schedule is to encourage helpful user behavior. Perhaps, without these incentives, those 20% of peers would not invest in larger capacity, too. And, if more users remained, the improvement would increase further.

In Figure 7,  $T_L$  is normalized to ESS as  $\Delta T_L$ , and  $\Delta T_L = (T_L - T_L^*) / T_L^*$ . Accordingly, the figure shows that ESS is a lower bound for  $\Delta T_L$ . In this case, the uRSF achieves the better  $\Delta T_L$  than eRSF and RSF, but the improvement is not significant. This is because uRSF can benefit more from faster RPs, and the supplementary capacities of the faster RPs are not substantial to have greater influence.

Looking at these figures, the results show that  $\Delta T_A$  of RSF is reduced significantly when the number of peers is about 60. The improvement reduces about 2/3 finish time of conventional ESS file delivery. Specifically,  $\Delta T_A$  is reduced down to 69.5%. Nevertheless,  $\Delta T_L$  continues to rise, and we observe a very interesting phenomenon, where RSF seems to have a scalability wall. This happens because the simulations have RSF-similar condition, and the continues rising  $\Delta T_L$  has greater impact on the performance. The lesser distributions of capacities, the lesser  $BW_k$  in the scheduling queue of RSF. To solve the problem, the peers could be divided into smaller size of clusters with RPs as server (*i.e.*, similar to super peers in Skype). And, if there are resources, service providers could offer a fast dummy peer to be the one who finish last, and use uRSF to offers better  $\Delta T_L$  (but  $\Delta T_A$  is slightly worse than eRSF). The optimal number of peers for  $BW_k$  is a novel insight deserving future study.

## 5. Conclusion

Our approach discretizes P2P scheduling, then solves each time interval with a min-min heuristic. Although the scheduler is sub-optimal, this is rarely a concern in practice, because: 1) it is more practical (P2P scheduling is NPcomplete where as RSF is fast), 2) close to optimal is as good as optimal, and 3) real-world networks will not entirely conform to our model, so optimizing it is not profound. Another group [15, 16, 17, 18, 19, 20] has developed an alternative scheduler, which mainly resulting from the tit-for-tat mechanism, for power saving, fairness, and robustness improvement of P2P networks. Their approaches do not include our innovations of partitioning and pathway restrictions.

We have also presented an adaptation that delivers good performance even if peers remain. Future work will extend this with a matrix of historical record to predict peer behavior and support full peer churn situation as well. Our scheduler is well-suited to this, because it operates in real-time – deciding on the allocation for  $t_{k+1}$  while the data of  $t_k$  is being sent. Any dynamic changes of system state can therefore be rolled into the schedule.

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