## A Proof of Theorem 1

Our proof has four main parts. In Appendix A.1, we bound the regret associated with the event that our high-probability confidence intervals do not hold. In Appendix A.2, we change counted events, from partially-observed suboptimal solutions to their fully-observed prefixes. In Appendix A.3, we bound the number of times that any suboptimal prefix can be chosen instead of the optimal solution *A*⇤. In Appendix A.4, we apply the counting argument of Kveton *et al.* [12] and finish our proof.

Let  $\mathbf{R}_t = R(\mathbf{A}_t, \mathbf{w}_t)$  be the stochastic regret of CombCascade at time t, where  $\mathbf{A}_t$  and  $\mathbf{w}_t$  are the solution and the weights of the items at time *t*, respectively. Let:

$$
\mathcal{E}_t = \left\{ \exists e \in E \text{ s.t. } \left| \bar{w}(e) - \hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) \right| \ge c_{t-1,\mathbf{T}_{t-1}(e)} \right\}
$$

be the event that  $\bar{w}(e)$  is outside of the high-probability confidence interval around  $\hat{w}_{\mathbf{T}_{t-1}(e)}(e)$  for at least one item  $e \in E$  at time *t*; and let  $\mathcal{E}_t$  be the complement of event  $\mathcal{E}_t$ , the event that  $\bar{w}(e)$  is in the high-probability confidence interval around  $\hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e)$  for all items  $e \in E$  at time *t*. Then we can decompose the expected regret of CombCascade as:

$$
R(n) = \mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\{\mathcal{E}_t\} \mathbf{R}_t\right] + \mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\{\overline{\mathcal{E}}_t\} \mathbf{R}_t\right].
$$
 (4)

#### A.1 Confidence Intervals Fail

The first term in (4) is easy to bound because  $\mathbf{R}_t$  is bounded and our confidence intervals hold with high probability. In particular, Hoeffding's inequality yields that for any *e*, *s*, and *t*:

$$
P(|\bar{w}(e) - \hat{\mathbf{w}}_s(e)| \geq c_{t,s}) \leq 2\exp[-3\log t],
$$

and therefore:

$$
\mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\{\mathcal{E}_{t}\}\right] \leq \sum_{e \in E} \sum_{t=1}^{n} \sum_{s=1}^{t} P(|\bar{w}(e) - \hat{\mathbf{w}}_{s}(e)| \geq c_{t,s})
$$
  

$$
\leq 2 \sum_{e \in E} \sum_{t=1}^{n} \sum_{s=1}^{t} \exp[-3\log t] \leq 2 \sum_{e \in E} \sum_{t=1}^{n} t^{-2} \leq \frac{\pi^{2}}{3}L.
$$

Since  $\mathbf{R}_t \leq 1$ ,  $\mathbb{E}\left[\sum_{t=1}^n \mathbb{1}\{\mathcal{E}_t\} \mathbf{R}_t\right] \leq \frac{\pi^2}{3}L$ .

#### A.2 From Partially-Observed Solutions to Fully-Observed Prefixes

Let  $\mathcal{H}_t = (\mathbf{A}_1, \mathbf{O}_1, \dots, \mathbf{A}_{t-1}, \mathbf{O}_{t-1}, \mathbf{A}_t)$  be the *history* of CombCascade up to choosing solution  $A_t$ , the first  $t-1$  observations and  $t$  actions. Let  $\mathbb{E}[\cdot|\mathcal{H}_t]$  be the conditional expectation given this history. Then we can rewrite the expected regret at time  $t$  conditioned on  $\mathcal{H}_t$  as:

$$
\mathbb{E}\left[\mathbf{R}_t\,|\,\mathcal{H}_t\right] = \mathbb{E}\left[\Delta_{\mathbf{A}_t}\mathbb{1}\{\Delta_{\mathbf{A}_t} > 0\} \,|\,\mathcal{H}_t\right] = \mathbb{E}\left[\frac{\Delta_{\mathbf{A}_t}}{p_{\mathbf{A}_t}}\mathbb{1}\{\Delta_{\mathbf{A}_t} > 0, \ \mathbf{O}_t \geq |\mathbf{A}_t|\}\,|\,\mathcal{H}_t\right]
$$

and analyze our problem under the assumption that all items in  $A_t$  are observed. This reduction is problematic because the probability  $p_{A_t}$  can be low, and as a result we get a loose regret bound. To address this problem, we introduce the notion of prefixes.

Let  $A = (a_1, \ldots, a_{|A|})$ . Then  $B = (a_1, \ldots, a_k)$  is a *prefix* of *A* for any  $k \leq |A|$ . In the rest of our analysis, we treat prefixes as feasible solutions to our original problem. Let  $B_t$  be a prefix of  $A_t$  as defined in Lemma 1. Then  $\Delta_{\mathbf{B}_t} \geq \frac{1}{2}\Delta_{\mathbf{A}_t}$  and  $p_{\mathbf{B}_t} \geq \frac{1}{2}f^*$ , and we can bound the expected regret at time *t* conditioned on  $\mathcal{H}_t$  as:

$$
\mathbb{E}\left[\mathbf{R}_{t} \mid \mathcal{H}_{t}\right] = \mathbb{E}\left[\frac{\Delta_{\mathbf{A}_{t}}}{p_{\mathbf{B}_{t}}}1\{\Delta_{\mathbf{A}_{t}} > 0, \mathbf{O}_{t} \geq |\mathbf{B}_{t}|\}\middle|\mathcal{H}_{t}\right]
$$

$$
\leq \frac{4}{f^{*}}\mathbb{E}\left[\Delta_{\mathbf{B}_{t}}1\{\Delta_{\mathbf{B}_{t}} > 0, \mathbf{O}_{t} \geq |\mathbf{B}_{t}|\}\middle|\mathcal{H}_{t}\right].
$$
(5)

Now we bound the second term in (4):

$$
\mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\left\{\overline{\mathcal{E}}_{t}\right\} \mathbf{R}_{t}\right] \stackrel{\text{(a)}}{=} \sum_{t=1}^{n} \mathbb{E}\left[\mathbb{1}\left\{\overline{\mathcal{E}}_{t}\right\} \mathbb{E}\left[\mathbf{R}_{t} \mid \mathcal{H}_{t}\right]\right] \n\stackrel{\text{(b)}}{\leq} \frac{4}{f^{*}} \mathbb{E}\left[\sum_{t=1}^{n} \Delta_{\mathbf{B}_{t}} \mathbb{1}\left\{\overline{\mathcal{E}}_{t}, \Delta_{\mathbf{B}_{t}} > 0, \mathbf{O}_{t} \geq |\mathbf{B}_{t}|\right\}\right].
$$
\n(6)

Equality (a) is due to the tower rule and that  $\mathbb{1}\{\overline{\mathcal{E}}_t\}$  is only a function of  $\mathcal{H}_t$ . Inequality (b) follows from the upper bound in (5).

### A.3 Counting Suboptimal Prefixes

Let:

$$
\mathcal{F}_t = \left\{ 2 \sum_{e \in \tilde{\mathbf{B}}_t} c_{n, \mathbf{T}_{t-1}(e)} \ge \Delta_{\mathbf{B}_t}, \ \Delta_{\mathbf{B}_t} > 0, \ \mathbf{O}_t \ge |\mathbf{B}_t| \right\} \tag{7}
$$

be the event that suboptimal prefix  $\mathbf{B}_t$  is "hard to distinguish" from  $A^*$ , where  $\tilde{\mathbf{B}}_t = \mathbf{B}_t \setminus A^*$  is the set of suboptimal items in  $\overline{B_t}$ . The goal of this section is to bound (6) by a function of  $\mathcal{F}_t$ .

We bound  $\Delta_{\mathbf{B}_t} 1\{\overline{\mathcal{E}}_t, \Delta_{\mathbf{B}_t} > 0, \mathbf{O}_t \geq |\mathbf{B}_t|\}$  from above for any suboptimal prefix  $\mathbf{B}_t$ . Our bound is proved based on several facts. First,  $\mathbf{B}_t$  is a prefix of  $\mathbf{A}_t$ , and hence  $f(\mathbf{B}_t, \mathbf{U}_t) \geq f(\mathbf{A}_t, \mathbf{U}_t)$  for any  $U_t$ . Second, when CombCascade chooses  $A_t$ ,  $f(A_t, U_t) \ge f(A^*, U_t)$ . It follows that:

$$
\prod_{e \in \mathbf{B}_t} \mathbf{U}_t(e) = f(\mathbf{B}_t, \mathbf{U}_t) \ge f(\mathbf{A}_t, \mathbf{U}_t) \ge f(A^*, \mathbf{U}_t) = \prod_{e \in A^*} \mathbf{U}_t(e).
$$

Now we divide both sides by  $\prod_{e \in A^* \cap B_t} U_t(e)$ :

$$
\prod_{e \in \tilde{\mathbf{B}}_t} \mathbf{U}_t(e) \ge \prod_{e \in A^* \setminus \mathbf{B}_t} \mathbf{U}_t(e)
$$

and substitute the definitions of the UCBs from (3):

$$
\prod_{e \in \tilde{\mathbf{B}}_t} \min \left\{ \hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) + c_{t-1,\mathbf{T}_{t-1}(e)}, 1 \right\} \ge \prod_{e \in A^* \backslash \mathbf{B}_t} \min \left\{ \hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) + c_{t-1,\mathbf{T}_{t-1}(e)}, 1 \right\}.
$$

Since  $\overline{\mathcal{E}}_t$  happens,  $|\overline{w}(e) - \hat{w}_{\mathbf{T}_{t-1}(e)}(e)| < c_{t-1,\mathbf{T}_{t-1}(e)}$  for all  $e \in E$  and therefore:

$$
\prod_{e\in A^*\backslash \mathbf{B}_t} \min\left\{\hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) + c_{t-1,\mathbf{T}_{t-1}(e)}, 1\right\} \ge \prod_{e\in A^*\backslash \mathbf{B}_t} \bar{w}(e)
$$
\n
$$
\prod_{e\in \tilde{\mathbf{B}}_t} \min\left\{\hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) + c_{t-1,\mathbf{T}_{t-1}(e)}, 1\right\} \le \prod_{e\in \tilde{\mathbf{B}}_t} \min\left\{\bar{w}(e) + 2c_{t-1,\mathbf{T}_{t-1}(e)}, 1\right\}.
$$

By Lemma 2:

$$
\prod_{e \in \tilde{\mathbf{B}}_t} \min \left\{ \bar{w}(e) + 2c_{t-1,\mathbf{T}_{t-1}(e)}, 1 \right\} \le \prod_{e \in \tilde{\mathbf{B}}_t} \bar{w}(e) + 2 \sum_{e \in \tilde{\mathbf{B}}_t} c_{t-1,\mathbf{T}_{t-1}(e)}.
$$

Finally, we chain the last four inequalities and get:

$$
\prod_{e \in \tilde{\mathbf{B}}_t} \bar{w}(e) + 2 \sum_{e \in \tilde{\mathbf{B}}_t} c_{t-1, \mathbf{T}_{t-1}(e)} \ge \prod_{e \in A^* \setminus \mathbf{B}_t} \bar{w}(e),
$$

which further implies that:

$$
2\sum_{e \in \tilde{\mathbf{B}}_t} c_{t-1,\mathbf{T}_{t-1}(e)} \ge \prod_{e \in A^* \backslash \mathbf{B}_t} \bar{w}(e) - \prod_{e \in \tilde{\mathbf{B}}_t} \bar{w}(e)
$$

$$
\ge \prod_{\substack{e \in A^* \cap \mathbf{B}_t}} \bar{w}(e) \left[ \prod_{e \in A^* \backslash \mathbf{B}_t} \bar{w}(e) - \prod_{e \in \tilde{\mathbf{B}}_t} \bar{w}(e) \right]
$$

$$
= \Delta_{\mathbf{B}_t}.
$$

Since  $c_{n,\mathbf{T}_{t-1}(e)} \ge c_{t-1,\mathbf{T}_{t-1}(e)}$  for any time  $t \le n$ , the event  $\mathcal{F}_t$  in (7) happens. Therefore, we can bound the right-hand side in (6) as:

$$
\mathbb{E}\left[\sum_{t=1}^n \Delta_{\mathbf{B}_t} \mathbb{1}\big\{\overline{\mathcal{E}}_t, \ \Delta_{\mathbf{B}_t} > 0, \ \mathbf{O}_t \geq |\mathbf{B}_t|\big\}\right] \leq \mathbb{E}\left[\hat{\mathbf{R}}(n)\right],
$$

where:

$$
\hat{\mathbf{R}}(n) = \sum_{t=1}^{n} \Delta_{\mathbf{B}_t} \mathbb{1} \{ \mathcal{F}_t \} . \tag{8}
$$

### A.4 CombUCB1 Analysis of Kveton *et al.* [12]

It remains to bound  $\hat{\mathbf{R}}(n)$  in (8). Note that the event  $\mathcal{F}_t$  can happen only if the weights of all items in  $B_t$  are observed. As a result,  $\hat{R}(n)$  can be bounded as in stochastic combinatorial semi-bandits. The key idea of our proof is to introduce infinitely-many mutually-exclusive events and then bound the number of times that these events happen when a suboptimal prefix is chosen [12]. The event *i* at time *t* is:

$$
G_{i,t} = \{\text{less than } \beta_1 K \text{ items in } \tilde{\mathbf{B}}_t \text{ were observed at most } \alpha_1 \frac{K^2}{\Delta_{\mathbf{B}_t}^2} \log n \text{ times,}
$$
  
...,  
less than  $\beta_{i-1} K$  items in  $\tilde{\mathbf{B}}_t$  were observed at most  $\alpha_{i-1} \frac{K^2}{\Delta_{\mathbf{B}_t}^2} \log n$  times,  
at least  $\beta_i K$  items in  $\tilde{\mathbf{B}}_t$  were observed at most  $\alpha_i \frac{K^2}{\Delta_{\mathbf{B}_t}^2} \log n$  times,

$$
\mathbf{O}_t \geq |\mathbf{B}_t|\},\
$$

where we assume that  $\Delta_{\mathbf{B}_t} > 0$ ; and the constants  $(\alpha_i)$  and  $(\beta_i)$  are defined as:

$$
1 = \beta_0 > \beta_1 > \beta_2 > \ldots > \beta_k > \ldots
$$

$$
\alpha_1 > \alpha_2 > \ldots > \alpha_k > \ldots,
$$

and satisfy  $\lim_{i\to\infty} \alpha_i = \lim_{i\to\infty} \beta_i = 0$ . By Lemma 3 of Kveton *et al.* [12],  $G_{i,t}$  are exhaustive at any time *t* when  $(\alpha_i)$  and  $(\beta_i)$  satisfy:

$$
\sqrt{6} \sum_{i=1}^{\infty} \frac{\beta_{i-1} - \beta_i}{\sqrt{\alpha_i}} \le 1.
$$

In this case:

$$
\widehat{\mathbf{R}}(n) = \sum_{t=1}^{n} \Delta_{\mathbf{B}_t} \mathbb{1}\{\mathcal{F}_t\} = \sum_{i=1}^{\infty} \sum_{t=1}^{n} \Delta_{\mathbf{B}_t} \mathbb{1}\{G_{i,t}, \Delta_{\mathbf{B}_t} > 0\}.
$$

Now we introduce item-specific variants of events  $G_{i,t}$  and associate the regret at time  $t$  with these events. In particular, let:

$$
G_{e,i,t} = G_{i,t} \cap \left\{ e \in \tilde{\mathbf{B}}_t, \ \mathbf{T}_{t-1}(e) \le \alpha_i \frac{K^2}{\Delta_{\mathbf{B}_t}} \log n \right\}
$$

be the event that item  $e$  is not observed "sufficiently often" under event  $G_{i,t}$ . Then it follows that:

$$
\mathbb{1}\{G_{i,t}, \Delta_{\mathbf{B}_t} > 0\} \le \frac{1}{\beta_i K} \sum_{e \in \tilde{E}} \mathbb{1}\{G_{e,i,t}, \Delta_{\mathbf{B}_t} > 0\}
$$

because at least  $\beta_i K$  items are not observed "sufficiently often" under event  $G_{i,t}$ . Therefore, we can bound  $\hat{\mathbf{R}}(n)$  as:

$$
\hat{\mathbf{R}}(n) \leq \sum_{e \in \tilde{E}} \sum_{i=1}^{\infty} \sum_{t=1}^{n} \mathbb{1} \{ G_{e,i,t}, \Delta_{\mathbf{B}_t} > 0 \} \frac{\Delta_{\mathbf{B}_t}}{\beta_i K}.
$$

Let each item *e* be in  $N_e$  suboptimal prefixes and  $\Delta_{e,1} \geq \ldots \geq \Delta_{e,N_e}$  be the gaps of these prefixes, ordered from the largest gap to the smallest. Then  $\hat{\mathbf{R}}(n)$  can be further bounded as:

$$
\hat{\mathbf{R}}(n) \leq \sum_{e \in \tilde{E}} \sum_{i=1}^{\infty} \sum_{t=1}^{n} \sum_{k=1}^{N_e} \mathbb{1} \{ G_{e,i,t}, \Delta_{\mathbf{B}_t} = \Delta_{e,k} \} \frac{\Delta_{e,k}}{\beta_i K}
$$
\n
$$
\leq \sum_{e \in \tilde{E}} \sum_{i=1}^{\infty} \sum_{t=1}^{n} \sum_{k=1}^{N_e} \mathbb{1} \left\{ e \in \tilde{\mathbf{B}}_t, \mathbf{T}_{t-1}(e) \leq \alpha_i \frac{K^2}{\Delta_{e,k}^2} \log n, \Delta_{\mathbf{B}_t} = \Delta_{e,k}, \mathbf{O}_t \geq |\mathbf{B}_t| \right\} \frac{\Delta_{e,k}}{\beta_i K}
$$
\n
$$
\leq \sum_{e \in \tilde{E}} \sum_{i=1}^{\infty} \frac{\alpha_i K \log n}{\beta_i} \left[ \Delta_{e,1} \frac{1}{\Delta_{e,1}^2} + \sum_{k=2}^{N_e} \Delta_{e,k} \left( \frac{1}{\Delta_{e,k}^2} - \frac{1}{\Delta_{e,k-1}^2} \right) \right]
$$
\n
$$
\leq \sum_{e \in \tilde{E}} \sum_{i=1}^{\infty} \frac{\alpha_i K \log n}{\beta_i} \frac{2}{\Delta_{e,N_e}}
$$
\n
$$
= \sum_{e \in \tilde{E}} K \frac{2}{\Delta_{e,N_e}} \left[ \sum_{i=1}^{\infty} \frac{\alpha_i}{\beta_i} \right] \log n,
$$

where inequality (a) follows from the definition of  $G_{e,i,t}$  and inequality (b) is from solving:

$$
\max_{A_{1:n},O_{1:n}} \sum_{t=1}^n \sum_{k=1}^{N_e} 1 \left\{ e \in \tilde{B}_t, T_{t-1}^{A_{1:n},O_{1:n}}(e) \le \alpha_i \frac{K^2}{\Delta_{e,k}^2} \log n, \ \Delta_{B_t} = \Delta_{e,k}, \ O_t \ge |B_t| \right\} \frac{\Delta_{e,k}}{\beta_i K},
$$

where  $A_{1:n} = (A_1, \ldots, A_n)$  is a sequence of *n* solutions,  $O_{1:n} = (O_1, \ldots, O_n)$  is a sequence of *n* observations,  $T_t^{A_{1:n},O_{1:n}}(e)$  is the number of times that item *e* is observed in *t* steps under  $A_{1:n}$  and  $O_{1:n}$ ,  $B_t$  is the prefix of  $A_t$  as defined in Lemma 1, and  $\tilde{B}_t = B_t \setminus A^*$ . Inequality (c) is by Lemma 3 of Kveton *et al.* [11]:

$$
\left[\Delta_{e,1} \frac{1}{\Delta_{e,1}^2} + \sum_{k=2}^{N_e} \Delta_{e,k} \left( \frac{1}{\Delta_{e,k}^2} - \frac{1}{\Delta_{e,k-1}^2} \right) \right] < \frac{2}{\Delta_{e,N_e}} \,.
$$

For the same  $(\alpha_i)$  and  $(\beta_i)$  as in Theorem 4 of Kveton *et al.* [12],  $\sum_{i=1}^{\infty} \frac{\alpha_i}{\beta_i} < 267$ . Moreover, since  $\Delta_{\mathbf{B}_t} \geq \frac{1}{2} \Delta_{\mathbf{A}_t}$  for any solution  $\mathbf{A}_t$  and its prefix  $\mathbf{B}_t$ , we have  $\Delta_{e,N_e} \geq \frac{1}{2} \Delta_{e,\text{min}}$ . Now we chain all inequalities and get:

$$
R(n) \leq \frac{4}{f^*} \mathbb{E}\left[\hat{\mathbf{R}}(n)\right] + \frac{\pi^2}{3} L \leq \frac{K}{f^*} \sum_{e \in \tilde{E}} \frac{4272}{\Delta_{e,\min}} \log n + \frac{\pi^2}{3} L.
$$

# B Proof of Theorem 2

The key idea is to decompose the regret of CombCascade into two parts, where the gaps  $\Delta_{A_t}$  are at most  $\epsilon$  and larger than  $\epsilon$ . In particular, note that for any  $\epsilon > 0$ :

$$
R(n) = \mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\{\Delta_{\mathbf{A}_t} \leq \varepsilon\} \mathbf{R}_t\right] + \mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\{\Delta_{\mathbf{A}_t} > \varepsilon\} \mathbf{R}_t\right].
$$
 (9)

The first term in (9) can be bounded trivially as:

$$
\mathbb{E}\left[\sum_{t=1}^{n} \mathbb{1}\{\Delta_{\mathbf{A}_t} \leq \varepsilon\} \mathbf{R}_t\right] = \mathbb{E}\left[\sum_{t=1}^{n} \Delta_{\mathbf{A}_t} \mathbb{1}\{\Delta_{\mathbf{A}_t} \leq \varepsilon, \Delta_{\mathbf{A}_t} > 0\}\right] \leq \epsilon n
$$

because  $\Delta_{\mathbf{A}_t} \leq \varepsilon$ . The second term in (9) can be bounded in the same way as  $R(n)$  in Theorem 1. The only difference is that  $\Delta_{e,\min} \geq \epsilon$  for all  $e \in \tilde{E}$ . Therefore:

$$
\mathbb{E}\left[\sum_{t=1}^n \mathbb{1}\{\Delta_{\mathbf{A}_t} > \varepsilon\} \mathbf{R}_t\right] \le \frac{K}{f^*} \sum_{e \in \tilde{E}} \frac{4272}{\Delta_{e,\min}} \log n + \frac{\pi^2}{3} L \le \frac{4272KL}{f^* \epsilon} \log n + \frac{\pi^2}{3} L.
$$

Now we chain all inequalities and get:

$$
R(n) \le \frac{4272KL}{f^*\epsilon} \log n + \epsilon n + \frac{\pi^2}{3}L.
$$
  
Finally, we choose  $\epsilon = \sqrt{\frac{4272KL\log n}{f^*n}}$  and get:  

$$
R(n) \le 2\sqrt{4272}\sqrt{\frac{KLn\log n}{f^*} + \frac{\pi^2}{3}L} < 131\sqrt{\frac{KLn\log n}{f^*} + \frac{\pi^2}{3}L},
$$

which concludes our proof.

# C Technical Lemmas

**Lemma 1.** Let  $A = (a_1, \ldots, a_{|A|}) \in \Theta$  be a feasible solution and  $B_k = (a_1, \ldots, a_k)$  be a prefix of  $k \leq |A|$  *items of A. Then*  $k$  *can be set such that*  $\Delta_{B_k} \geq \frac{1}{2}\Delta_A$  *and*  $p_{B_k} \geq \frac{1}{2}f^*$ *.* 

*Proof.* We consider two cases. First, suppose that  $f(A, \bar{w}) \geq \frac{1}{2} f^*$ . Then our claims hold trivially for  $k = |A|$ . Now suppose that  $f(A, \bar{w}) < \frac{1}{2} f^*$ . Then we choose  $k$  such that:

$$
f(B_k, \bar{w}) \leq \frac{1}{2} f^* \leq p_{B_k} .
$$

Such *k* is guaranteed to exist because  $\bigcup_{i=1}^{|A|} [f(B_i, \bar{w}), p_{B_i}] = [f(A, \bar{w}), 1]$ , which follows from the facts that  $f(B_i, \bar{w}) = p_{B_i} \bar{w}(a_i)$  for any  $i \le |A|$  and  $p_{B_1} = 1$ . We prove that  $\Delta_{B_k} \ge \frac{1}{2} \Delta_A$  as:

$$
\Delta_{B_k} = f^* - f(B_k, \bar{w}) \ge \frac{1}{2} f^* \ge \frac{1}{2} \Delta_A.
$$

The first inequality is by our assumption and the second one holds for any solution *A*.

**Lemma 2.** *Let*  $0 \leq p_1, ..., p_K \leq 1$  *and*  $u_1, ..., u_K \geq 0$ *. Then:* 

$$
\prod_{k=1}^{K} \min \{p_k + u_k, 1\} \le \prod_{k=1}^{K} p_k + \sum_{k=1}^{K} u_k.
$$

*This bound is tight when*  $p_1, ..., p_K = 1$  *and*  $u_1, ..., u_K = 0$ .

*Proof.* The proof is by induction on *K*. Our claim clearly holds when  $K = 1$ . Now choose  $K > 1$ and suppose that our claim holds for any  $0 \leq p_1, \ldots, p_{K-1} \leq 1$  and  $u_1, \ldots, u_{K-1} \geq 0$ . Then:

$$
\prod_{k=1}^{K} \min \{p_k + u_k, 1\} = \min \{p_K + u_K, 1\} \prod_{k=1}^{K-1} \min \{p_k + u_k, 1\}
$$
\n
$$
\leq \min \{p_K + u_K, 1\} \left(\prod_{k=1}^{K-1} p_k + \sum_{k=1}^{K-1} u_k\right)
$$
\n
$$
\leq p_K \prod_{k=1}^{K-1} p_k + u_K \prod_{\substack{k=1 \le k \le K}}^{K-1} p_k + \min \{p_K + u_K, 1\} \sum_{\substack{k=1 \le k \le K}}^{K-1} u_k
$$
\n
$$
\leq \prod_{k=1}^{K} p_k + \sum_{k=1}^{K} u_k.
$$

14