

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:

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

be the event that $\bar{w}(e)$ is outside of the high-probability confidence interval around $\hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e)$ for at least one item $e \in E$ at time t ; and let $\bar{\mathcal{E}}_t$ 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:

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

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:

$$\begin{aligned} \mathbb{E} \left[\sum_{t=1}^n \mathbb{1}\{\mathcal{E}_t\} \right] &\leq \sum_{e \in E} \sum_{t=1}^n \sum_{s=1}^t P(|\bar{w}(e) - \hat{\mathbf{w}}_s(e)| \geq c_{t,s}) \\ &\leq 2 \sum_{e \in E} \sum_{t=1}^n \sum_{s=1}^t \exp[-3 \log t] \leq 2 \sum_{e \in E} \sum_{t=1}^n t^{-2} \leq \frac{\pi^2}{3} L. \end{aligned}$$

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

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 \mathbf{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:

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

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

Let $A = (a_1, \dots, a_{|A|})$. Then $B = (a_1, \dots, a_k)$ 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 \mathbf{B}_t be a prefix of \mathbf{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:

$$\begin{aligned} \mathbb{E}[\mathbf{R}_t | \mathcal{H}_t] &= \mathbb{E} \left[\frac{\Delta_{\mathbf{A}_t}}{p_{\mathbf{A}_t}} \mathbb{1}\{\Delta_{\mathbf{A}_t} > 0, \mathbf{O}_t \geq |\mathbf{A}_t|\} \middle| \mathcal{H}_t \right] \\ &\leq \frac{4}{f^*} \mathbb{E}[\Delta_{\mathbf{B}_t} \mathbb{1}\{\Delta_{\mathbf{B}_t} > 0, \mathbf{O}_t \geq |\mathbf{B}_t|\} | \mathcal{H}_t]. \end{aligned} \quad (5)$$

Now we bound the second term in (4):

$$\begin{aligned} \mathbb{E} \left[\sum_{t=1}^n \mathbb{1}\{\bar{\mathcal{E}}_t\} \mathbf{R}_t \right] &\stackrel{(a)}{=} \sum_{t=1}^n \mathbb{E} [\mathbb{1}\{\bar{\mathcal{E}}_t\} \mathbb{E} [\mathbf{R}_t | \mathcal{H}_t]] \\ &\stackrel{(b)}{\leq} \frac{4}{f^*} \mathbb{E} \left[\sum_{t=1}^n \Delta_{\mathbf{B}_t} \mathbb{1}\{\bar{\mathcal{E}}_t, \Delta_{\mathbf{B}_t} > 0, \mathbf{O}_t \geq |\mathbf{B}_t|\} \right]. \end{aligned} \quad (6)$$

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

A.3 Counting Suboptimal Prefixes

Let:

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

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 \mathbf{B}_t . The goal of this section is to bound (6) by a function of \mathcal{F}_t .

We bound $\Delta_{\mathbf{B}_t} \mathbb{1}\{\bar{\mathcal{E}}_t, \Delta_{\mathbf{B}_t} > 0, \mathbf{O}_t \geq |\mathbf{B}_t|\}$ 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 \mathbf{U}_t . Second, when CombCascade chooses \mathbf{A}_t , $f(\mathbf{A}_t, \mathbf{U}_t) \geq f(A^*, \mathbf{U}_t)$. It follows that:

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

Now we divide both sides by $\prod_{e \in A^* \cap \mathbf{B}_t} \mathbf{U}_t(e)$:

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

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

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

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

$$\begin{aligned} \prod_{e \in A^* \setminus \mathbf{B}_t} \min \{ \hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) + c_{t-1, \mathbf{T}_{t-1}(e)}, 1 \} &\geq \prod_{e \in A^* \setminus \mathbf{B}_t} \bar{w}(e) \\ \prod_{e \in \tilde{\mathbf{B}}_t} \min \{ \hat{\mathbf{w}}_{\mathbf{T}_{t-1}(e)}(e) + c_{t-1, \mathbf{T}_{t-1}(e)}, 1 \} &\leq \prod_{e \in \tilde{\mathbf{B}}_t} \min \{ \bar{w}(e) + 2c_{t-1, \mathbf{T}_{t-1}(e)}, 1 \}. \end{aligned}$$

By Lemma 2:

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

Finally, we chain the last four inequalities and get:

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

which further implies that:

$$\begin{aligned} 2 \sum_{e \in \tilde{\mathbf{B}}_t} c_{t-1, \mathbf{T}_{t-1}(e)} &\geq \prod_{e \in A^* \setminus \mathbf{B}_t} \bar{w}(e) - \prod_{e \in \tilde{\mathbf{B}}_t} \bar{w}(e) \\ &\geq \underbrace{\prod_{e \in A^* \cap \mathbf{B}_t} \bar{w}(e)}_{\leq 1} \left[\prod_{e \in A^* \setminus \mathbf{B}_t} \bar{w}(e) - \prod_{e \in \tilde{\mathbf{B}}_t} \bar{w}(e) \right] \\ &= \Delta_{\mathbf{B}_t}. \end{aligned}$$

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

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

where:

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

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 \mathbf{B}_t are observed. As a result, $\hat{\mathbf{R}}(n)$ 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:

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

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

$$\begin{aligned} 1 = \beta_0 > \beta_1 > \beta_2 > \dots > \beta_k > \dots \\ \alpha_1 > \alpha_2 > \dots > \alpha_k > \dots , \end{aligned}$$

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

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

In this case:

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

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

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

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

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

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 $\Delta_{e,1} \geq \dots \geq \Delta_{e,N_e}$ be the gaps of these prefixes, ordered from the largest gap to the smallest. Then $\hat{\mathbf{R}}(n)$ can be further bounded as:

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

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

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

where $A_{1:n} = (A_1, \dots, A_n)$ is a sequence of n solutions, $O_{1:n} = (O_1, \dots, O_n)$ 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 (α_i) and (β_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 $\Delta_{e,N_e} \geq \frac{1}{2} \Delta_{e,\min}$. Now we chain all inequalities and get:

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

B Proof of Theorem 2

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

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

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

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

because $\Delta_{\mathbf{A}_t} \leq \epsilon$. 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:

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

Now we chain all inequalities and get:

$$R(n) \leq \frac{4272KL}{f^* \epsilon} \log n + \epsilon n + \frac{\pi^2}{3} L.$$

Finally, we choose $\epsilon = \sqrt{\frac{4272KL \log n}{f^* n}}$ and get:

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

which concludes our proof.

C Technical Lemmas

Lemma 1. *Let $A = (a_1, \dots, a_{|A|}) \in \Theta$ be a feasible solution and $B_k = (a_1, \dots, a_k)$ 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 $i \leq |A|$ and $p_{B_1} = 1$. We prove that $\Delta_{B_k} \geq \frac{1}{2} \Delta_A$ as:

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

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

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

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

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

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 $0 \leq p_1, \dots, p_{K-1} \leq 1$ and $u_1, \dots, u_{K-1} \geq 0$. Then:

$$\begin{aligned} \prod_{k=1}^K \min\{p_k + u_k, 1\} &= \min\{p_K + u_K, 1\} \prod_{k=1}^{K-1} \min\{p_k + u_k, 1\} \\ &\leq \min\{p_K + u_K, 1\} \left(\prod_{k=1}^{K-1} p_k + \sum_{k=1}^{K-1} u_k \right) \\ &\leq p_K \prod_{k=1}^{K-1} p_k + u_K \underbrace{\prod_{k=1}^{K-1} p_k}_{\leq 1} + \underbrace{\min\{p_K + u_K, 1\}}_{\leq 1} \sum_{k=1}^{K-1} u_k \\ &\leq \prod_{k=1}^K p_k + \sum_{k=1}^K u_k. \end{aligned}$$

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