

Prophet Secretary

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Given that F_1, F_2, \dots, F_n are the initial distribution and V_1, V_2, \dots, V_n are the corresponding sampled values. For any $t \in [0, 1]$, define:

$\theta(t)$ = The probability that nothing is selected till time t .

$q_{-i}(t)$ = The probability that nothing is selected till time t conditioned that V_i arrives at time t .

$\alpha : [0, 1] \rightarrow [0, 1]$ and $\beta : [0, 1] \rightarrow \mathbb{R}$, both non-increasing such that,

$$\Pr[\max_i \{V_1, V_2, \dots, V_n\} \leq \alpha(t)] = \beta(t)$$

Lemma 0.1. $\forall i \in [n], t \in [0, 1]$

$$q_{-i}(t) = \frac{\theta(t)}{1 - t + \int_0^t \Pr[V_i \leq \beta(x)] dx}$$

Proof.

$$\begin{aligned} \theta(t) &= \int_0^1 \Pr[T > t \mid t_i = x] dx \\ &= \int_0^t \Pr[T > t \mid t_i = x] dx + \int_t^1 \Pr[T > t \mid t_i = x] dx \end{aligned}$$

For any x greater than t ,

$$\Pr[T > t \mid t_i = x] = \Pr[T > t \mid t_i > t] = q_{-i}(t)$$

For any x less than t ,

$$\begin{aligned} \Pr[T > t \mid t_i = x] &= \Pr[V_i \leq \beta(x)] (\Pr[\text{Nothing is chosen before } x \mid V_i \text{ does not come before } x] \\ &\quad + \Pr[\text{Nothing is chosen from } x \text{ to } t \mid V_i \text{ does not come between } x \text{ and } t]) \\ &= \Pr[V_i \leq \beta(x)] (\Pr[\text{Nothing is chosen before } t \text{ and } V_i \text{ does not come before } t]) \\ &= \Pr[V_i \leq \beta(x)] (q_{-i}(t)) \end{aligned}$$

Putting the inequalities and rearranging gives the result. □

Lemma 0.2. $\forall t \in [0, 1/2]$

$$\sum_{i=1}^n \frac{\Pr[V_i > \beta(t)]}{1 - \int_0^t \Pr[V_i > \beta(x)] dx} \geq \frac{\Pr[\max_i \{V_i\} > \beta(t)]}{1 - t \Pr[\max_i \{V_i\} > \beta(0)]}$$

Proof.

$$\begin{aligned}
\sum_{i=1}^n \frac{\Pr[V_i > \beta(t)]}{1 - \int_0^t \Pr[V_i > \beta(x)]dx} &= \sum_{i=1}^n \frac{\Pr[V_i > \beta(t)]}{1 - t + \int_0^t \Pr[V_i \leq \beta(x)]dx} \\
&\geq \sum_{i=1}^n \frac{\Pr[V_i > \beta(t)]}{1 - t + t \Pr[V_i \leq \beta(0)]} \\
&= \sum_{i=1}^n \frac{1 - F_i(\beta(t))}{1 - t + t F_i(\beta(0))}
\end{aligned}$$

Using the following inequality (Proof in the paper)

$$\frac{1 - F_1(\beta(t))}{1 - t + t F_1(\beta(0))} + \frac{1 - F_2(\beta(t))}{1 - t + t F_2(\beta(0))} \geq \frac{1 - F_1(\beta(t)) F_2(\beta(t))}{1 - t + t F_1(\beta(0)) F_2(\beta(0))}$$

and repeating it n times, we get the required result. \square

For the competitive ratio of 0.669, we define $\alpha_1, \alpha_2, \dots, \alpha_m$ and set $\alpha(t) = \alpha_{\text{ceil}(tm)}$. We also define a non-decreasing function g , for $t \in [0, 1]$,

$$g_p(t) = \begin{cases} \frac{1}{1-t(1-p)} & ; t \leq 1/2 \\ \frac{2}{1+p} & ; t > 1/2 \end{cases}$$

Using the result from the last section, if z is the reward, for any $t \in [0, 1]$, we can write,

$$\Pr[z > \beta(t)] = \frac{1 - \theta(t)}{1 - \alpha(t)} (1 - \alpha(t)) + \sum_{i \in [n]} \Pr[V_i > \beta(t)] \int_t^1 q_{-1}(x) dx$$

Using the bound for $\theta(t)$ derived in the last part, the first term can be written as,

$$\frac{1 - \theta(t)}{1 - \alpha(t)} (1 - \alpha(t)) \geq \frac{\int_0^t 1 - \alpha(x) dx}{1 - \alpha(t)} (\Pr[\max_i \{V_i\} > \beta_t])$$

For the second term, first we use Lemma 0.1 to write,

$$q_{-i}(x) \geq \frac{\theta(x)}{1 - \int_0^x \Pr[V_i > \beta(y)] dy}$$

Then interchanging the order of sums,

$$\sum_{i \in [n]} \Pr[V_i > \beta(t)] \int_t^1 q_{-1}(x) dx \geq \int_t^1 \theta(x) \sum_{i \in [n]} \frac{\Pr[V_i > \beta(t)]}{1 - \int_0^x \Pr[V_i > \beta(y)] dy} dx$$

Now, using Lemma 2, for $x \leq 1/2$

$$\begin{aligned}
\sum_{i \in [n]} \frac{\Pr[V_i > \beta(t)]}{1 - \int_0^x \Pr[V_i > \beta(y)] dy} &\geq \frac{\Pr[\max_i \{V_i\} > \beta(x)]}{1 - x \Pr[\max_i \{V_i\} > \beta(0)]} \\
&= g_{\alpha(0)}(x) \Pr[\max_i \{V_i\} > \beta(x)]
\end{aligned}$$

Now, for $x > 1/2$, it can be observed that

$$\begin{aligned} \sum_{i \in [n]} \frac{\Pr[V_i > \beta(t)]}{1 - \int_0^x \Pr[V_i > \beta(y)] dy} &\geq \frac{\Pr[\max_i \{V_i\} > \beta(x)]}{1 - \frac{1}{2} \Pr[\max_i \{V_i\} > \beta(0)]} \\ &= g_{\alpha(0)}(x) \Pr[\max_i \{V_i\} > \beta(x)] \end{aligned}$$

Finally, we divide x according to the m thresholds, and use the bound for $\theta(t)$ derived in the last section to get the final result.

$$\begin{aligned} \int_t^1 \theta(x) g_{\alpha(0)}(x) dx &= \sum_{i=tm}^m \int_{i/m}^{(i+1)/m} \theta(x) g_{\alpha(0)}(x) dx \\ &= \sum_{i=tm}^m \int_{i/m}^{(i+1)/m} \theta(i/m) \Pr[V_{time=x} < \beta_i]^{x-i/m} g_{\alpha(0)}(x) dx \\ &\geq \sum_{i=tm}^m \int_{i/m}^{(i+1)/m} \theta(i/m) \Pr[V_{time=x} < \beta_i]^{x-i/m} g_{\alpha(0)}(i/m) dx \\ &\geq \sum_{i=tm}^m g_{\alpha(0)}(i/m) e^{\int_0^{i/m} \ln(\alpha(y)) dy} \int_{i/m}^{(i+1)/m} \alpha_i^{x-i/m} dx \\ &= \sum_{i=tm}^m g_{\alpha(0)}(i/m) \left(\prod_{k=1}^i \alpha_k \right)^{1/m} \int_{i/m}^{(i+1)/m} \alpha_i^{x-i/m} dx \\ &= \sum_{i=tm}^m g_{\alpha(0)}(i/m) \left(\prod_{k=1}^i \alpha_k \right)^{1/m} \frac{1 - \alpha_i^{1/m}}{-\ln(\alpha_i)} \end{aligned}$$

Hence, we proved that, for each $t \in [0, 1]$:

$$\Pr[z > \beta(t)] \geq \min_t \left(\frac{\int_0^t 1 - \alpha(x) dx}{1 - \alpha(t)} + \sum_{i=tm}^m g_{\alpha(0)}(i/m) \left(\prod_{k=1}^i \alpha_k \right)^{1/m} \frac{1 - \alpha_i^{1/m}}{-\ln(\alpha_i)} \right) \Pr[\max_i \{V_i\} > \beta_t]$$

Where optimization of values for α lead to a competitive ratio of 0.669