

# Prophet Inequalities with Oracle Calls

and why they're useful

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Observe realizations  $X_1, X_2, \dots, X_n \sim (\text{known}) \mathcal{D}_1, \mathcal{D}_2, \dots, \mathcal{D}_n$ .

- ▶ Order is *adversarial*.
- ▶ Design algorithm to maximize selected value.
- ▶ Compare against all-knowing *prophet*.
- ▶ If  $\mathcal{D}_1 = \mathcal{D}_2 = \dots = \mathcal{D}_n \implies$  IID setting.

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- ▶ Competitive Ratio (CR or “Prophet Objective”):

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$$\frac{\mathbb{E}[ALG]}{\mathbb{E}[\max_i X_i]}$$

- ▶ Probability of Selecting Maximum Realization (PMax or “Secretary Objective”):

$$\Pr \left[ ALG \text{ selects } \max_i X_i \right]$$

# What is known?

Prophet Objective [Krengel, Sucheston, Garling '77, '78]

$\exists$  stopping strategy that achieves  $1/2 \cdot \mathbb{E}[\max_i X_i]$ , and this is tight.

Secretary Objective

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Motivation: Auctions/Hiring with *overbooking*.

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Q.: Is  $\mathcal{O}_k$  equivalent to Top-1-of- $(k+1)$ ?

A.: YES and NO.

# Let's Play!

$$k = 1$$

$$\mathcal{U}[13, 14]$$

$$\mathcal{U}[7, 16]$$

$$\mathcal{U}[0, 20]$$

$$\begin{cases} 1000 & \text{w.p. } \frac{1}{100} \\ 0 & \text{otherwise} \end{cases}$$

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$$X_1 = 13.67$$

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$$X_4 = 0$$

$$\mathbb{E}[\max \{X_1, X_2, X_3, X_4\}] \approx 24.66$$

$$\mathbb{E}[\text{OPTALG} \{X_1, X_2, X_3, X_4\} \text{ for } k = 0] \approx 13.37$$

$$\mathbb{E}[\text{OPTALG} \{X_1, X_2, X_3, X_4\} \text{ for } k = 1] \approx 19.9$$

Optimal strategy was to ignore  $X_1, X_2$  and query at  $X_3$ .

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$\mathcal{O}_k \not\equiv \text{Top-1-of-}(k+1)$  for CR

Guarantee  $\alpha$  for  $\mathcal{O}_k \implies \alpha$  for Top-1-of- $(k+1)$ .

## $\mathcal{O}_k \not\equiv \text{Top-1-of-}(k+1)$ for CR

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Fix  $\varepsilon > 0$ ,  $k = 1$ , and let

$$X_1 = 1, \quad X_2 = \begin{cases} 1 + \varepsilon & \text{w.p. } 1/2 - \varepsilon \\ 0 & \text{w.p. } 1/2 + \varepsilon \end{cases}, \quad X_3 = \begin{cases} 1/\varepsilon & \text{w.p. } \varepsilon \\ 0 & \text{w.p. } 1 - \varepsilon \end{cases}$$

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- ▶  $\mathbb{E}[\max\{X_1, X_2, X_3\}] \rightarrow 2$ .
- ▶ Let  $\mathcal{A}$  for Top-1-of-2 that always selects  $X_1$  and  $X_3$ .  
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- ▶ Let  $\mathcal{B}$  for  $\mathcal{O}_1$ .
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► If  $\mathcal{B}$  queries at  $X_1 \implies$

$$\mathbb{E}[\mathcal{B}] = \left(\frac{1}{2} + \varepsilon\right) (1 - \varepsilon) \cdot 1 + \left(\frac{1}{2} - \varepsilon\right) \cdot (1 + \varepsilon) + \left(\frac{1}{2} + \varepsilon\right) \varepsilon \cdot \frac{1}{\varepsilon} \rightarrow 3/2.$$



# PMax for IID

Theorem [Har-Peled, Harb, L. '24]

Single-threshold\* algorithm  $\mathcal{A}$  achieves

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- ▶ Almost tight bound, up to the exponent.
- ▶ New upper bound.  
Asymptotical improvement on lower bound of  
[Gilbert, Mosteller '66]

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- ▶ Better idea: Set threshold  $T$  such that  $\Pr[X_i \geq T] = \frac{e^{\sqrt{k}}}{n}$ .

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▶ Upper Bound:  $\mathcal{U}[0, 1]$  with roughly  $k \log k$  of the  $X_i$ 's in  $[1 - k \log k / n, 1]$ .

Every algorithm will either miss  $\max_i X_i$  or run out of oracle calls w.p.  $k^{-k}$ .

# CR for Non-IID

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$\exists$  sequence  $\{\xi_k\}_{k \in \mathbb{N}}$  and single-threshold\* algorithm  $\mathcal{A}$  such that

$$\text{CR}(\mathcal{A}) \geq 1 - O(e^{-\xi_k}) = 1 - O(e^{-k/e + o(k)}).$$

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- ▶ Tight bound for the oracle model.
- ▶ Asymptotical improvement on both upper and lower bounds of [Ezra, Feldman, Nehama '18].

# CR for Non-IID

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$$\blacktriangleright 1 - e^{-\xi_k} = e^{-\xi_k} \sum_{i=0}^k \frac{(\xi_k)^i}{i!} \iff e^{-\xi_k} = e^{-\xi_k} \sum_{i=k+1}^{\infty} \frac{(\xi_k)^i}{i!}$$

$\xi_k$  : Exponent sequence.

Intuition: Balances  $\Pr[\text{Poi}(\xi_k) \geq k+1] = \Pr[\text{Poi}(\xi_k) = 0]$ .

# CR for Non-IID

$$\blacktriangleright 1 - e^{-\xi_k} = e^{-\xi_k} \sum_{i=0}^k \frac{(\xi_k)^i}{i!} \iff e^{-\xi_k} = e^{-\xi_k} \sum_{i=k+1}^{\infty} \frac{(\xi_k)^i}{i!}$$

$\xi_k$  : Exponent sequence.

Intuition: Balances  $\Pr[\text{Poi}(\xi_k) \geq k+1] = \Pr[\text{Poi}(\xi_k) = 0]$ .

$\blacktriangleright$  Idea: Set threshold  $T$  such that  $\Pr[\max_i X_i \geq T] = 1 - e^{-\xi_k}$ .

For  $x \geq T$ , analyze  $\Pr[ALG \geq x]$  using Poissonization technique of [Harb '23].



# CR for Non-IID

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For  $x \geq T$ , analyze  $\Pr[ALG \geq x]$  using Poissonization technique of [Harb '23].

*Like in standard PI, we set  $\Pr[\max_i X_i \geq T] = p$  and get a CR of  $p$ .*

► Upper Bound:

$$X_1 = 1, \quad X_2, \dots, X_{n-1} \sim \text{Poi}(\xi_k), \quad X_n = \begin{cases} 1/\varepsilon & \text{w.p. } \varepsilon \\ 0 & \text{w.p. } 1 - \varepsilon \end{cases}$$

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►  $\mathbb{E}[\max_i X_i] = 2.$

$$\begin{aligned} \mathbb{E}[ALG] &= 2 \cdot \Pr[\leq k \text{ of } X_2, \dots, X_{n-1} \neq 0] \\ &\quad + 1 \cdot (1 - \Pr[\leq k \text{ of } X_2, \dots, X_{n-1} \neq 0]) \\ &= 2(1 - e^{-\xi_k}), \end{aligned}$$

by the choice of  $\xi_k$ .

# Open Questions

## Top-1-of- $k$ Model:

- ▶ Optimal asymptotics for PMax in IID setting and CR in Non-IID setting (requires new approach).
- ▶ PMax in Non-IID?  
Analyzing oracle model is enough!
- ▶ CR in IID?

## Oracle Model:

- ▶ Incorrect predictions  $\implies$  Robustness, ties with ML.

# Questions?

