# Towards a Unified Theory of IID Prophet Inequalities

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Partly joint work with



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doesn't make any of the "end-use" equip ment mentioned below. Yet, some of these applications of Sorensen equipment by our stomers are so novel that they may be of interest to you. Maybe they'll spark an idea: Open Sesame, Selection of sesame seeds for use in the manufacture of halvah-a favorite confection of New York's lower East Side - was the job of kilevolts from one Sorensen Series 200 supply. Same principle can purify other grains and cereals, tobacco,

and low-grade ores. Gold From Air, Gold spun off into thin air from a grinding or buffing wheel can quickly cause cash to vanish. Ditto with platinum or other precious metals. Clever Soremen customers are putting this pay dirt back into the pay roll with an electrostatic recovery otherwise you may find the road to a system-powered, of course, with a Screenen solution blocked by an unwarranted as-

Ignition damper. Everybody's heard about the high-voltage spark that sets off an explosion. A new h-v system prevents explosions. High-voltage-from a Scrensen 9000 Series-precipitates a sample of potentially explosive dusts; an alarm is given long before the concentration becomes dangerous. Vanishing Volt-Amps. Dielectric testing with a-c is more or less standard. (Sorensen offers a complete line of hay are testers conforming to ASTM standards.) However, where the test load has high caracitance, d-c testing can often effect substantial savings. In a typical problem, a 250-watt, d-c tester replaced a 25 kva a-c tester with equal results, one-fourth the cost, and a 100:1 reduction in light bills.

High-voltage or low, you'll find that Sorensen has the answer to your controlled power problems. In addition to high-voltage equipand unregulated d-c supplies, a-c line-voltage regulators, frequency changers, inverters, Richards Ave., South Norwalk, Conn. o.es



... the widest line lets you make the wisest choice

IN 1960 SCIENTIFIC AMERICAN INC

#### MATHEMATICAL GAMES

A fifth collection of "brain-teasers"

by Martin Gardner

wery eight months or so this deof short problems drawn from various mathematical fields. This is the fifth such collection. The answers to the problems will be given here next month. welcome letters from readers who find fault with an answer, solve a problem more elegantly, or generalize a problem in some interesting way. In the past I have tried to avoid puzzles that play verbal pranks on the reader, so I think it only fair to say that several of this month's "brain-teasers" are touched with whimsy. They must be read with care:

Mel Stover of Winnipeg was the first to send this amusing problem-amusing because of the ease with which even the best of recometers may fail to approach it properly. Given a triangle with one abtuse angle, is it possible to cut the triangle into smaller triangles, all of them acute? (An acute triangle is a riangle with three acute angles. A right ingle is of course neither acute nor obtose ). If this cannot be done give a proof of impossibility. If it can be done, what is the smallest number of acute triangles into which any obtuse triangle

The illustration at right shows a typical attempt that leads nowhere. The triangle has been divided into three acute triangles, but the fourth is obtuse, so nothing has been gained by the pre-

This delightful problem led me to ask myself: "What is the smallest number of cute triangles into which a square can be dissected?" For days I was convinced that nine was the answer; then suddenly I saw how to reduce it to eight, I won der how many readers can discover an eight-triangle solution, or perhans an even better one. I am unable to prove that eight is the minimum, though I strongly suspect that it is.

In H. G. Wells's newel The First Men in the Moon our natural satellite is found to be inhabited by intelligent insect surface. These creatures, let us assure have a unit of distance that we shall call a "lenar." It was adopted because the moon's surface area, if expressed in square lunars, exactly equals the moon's volume in cubic lunars. The moon's diameter is 2,160 miles. How many miles long is a lunar?

In 1958 John H. Fox, Jr., of the Min

neurolis-Honorovell Retulator Co. and L. Gerald Marnie of the Massachusetts Institute of Technology devised an unusual betting game which they call Googol. It is played as follows: Ask someone to take as many slips of paper as he pleases, and on each slip write a different positive number. The numbers may range from small fractions of one to a number the size of a "googol" (1 follarger. These slips are turned face-down and shuffled over the top of a table. One at a time you turn the slins face on. The aim is to stop turning when you come to the number that you guess to be the largest of the series. You cannot go back and pick a previously turned slip. If you turn over all the slips, then of course you must pick the last one turned. Most people will suppose the odds

#### Secretary Problem

n unknown values

 $X_1, \ldots, X_n$ 

Random order

**▶** Step *i*:

1. Select  $x_i$  and stop

2. Ignore  $x_i$  and continue

 $Pr[We select max_i x_i]$ ?





.. •







 $S_1$  Sampling Phase

$$S_2$$
 Selection Phase

w.p. 
$$\frac{1}{2}$$
,  $x_1^* \in S_2$   
w.p.  $\frac{1}{2}$ ,  $x_2^* \in S_1$   $\Longrightarrow Pr[\text{We select max } x_i] \ge \frac{1}{4}$ .

1



.. •



Sampling Phase

Selection Phase

$$\left. \begin{array}{ll} \text{w.p. } 1/2, & x_1^* \in \mathcal{S}_2 \\ \text{w.p. } 1/2, & x_2^* \in \mathcal{S}_1 \end{array} \right\} \implies Pr[\text{We select} \max_i x_i] \geq 1/4.$$

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(-2)





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Sampling Phase

$$S_2$$
 Selection Phase

w.p. 
$$\frac{1}{2}$$
,  $x_1^* \in S_2$   
w.p.  $\frac{1}{2}$ ,  $x_2^* \in S_1$   $\Longrightarrow Pr[\text{We select max } x_i] \ge \frac{1}{4}$ .









Sampling Phase

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▶ Can get 1/e (optimal) by sampling first n/e.

## Prophet Inequality

What if we know *something* about the  $x_i$ 's? [Krengel, Sucheston and Garling '77]

 $X_1, X_2, \ldots, X_n \sim (\text{known}) \ \mathcal{D}_1, \mathcal{D}_2, \ldots, \mathcal{D}_n$  arrive in *adversarial* order.

- Design stopping time to maximize selected value.
- ▶ Compare against all-knowing *prophet*:  $\mathbb{E}[\max_i X_i]$ .

 $\mathcal{U}[2,4]$   $\mathcal{U}[2,4]$ 



 $\mathcal{U}[0,7]$ 

 $\mathcal{U}[2,4]$   $X_1 = 2.34$ 







 $\mathcal{U}[2,4]$ 

 $\mathcal{U}[2,4]$ 

 $\mathcal{U}[1,5]$ 

 $\mathcal{U}[0,7]$ 

 $X_1 = 2.34$ 

 $X_2 = 3.12$ 

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$$X_1 = 2.34$$

$$X_2 = 3.12$$

 $X_3 = 3.20$ 

 $X_4 = 0.87$ 

## Prophet Inequality [Krengel, Sucheston and Garling '77, '78]

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

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

 $\mathbb{E}\left[\mathsf{ALG}\right] = 1$  for all algorithms.

$$\mathbb{E}[\max_i X_i] = \frac{1}{\varepsilon} \cdot \varepsilon + 1 \cdot (1 - \varepsilon) = 2 - \varepsilon.$$

## Prophet Inequality [Krengel, Sucheston and Garling '77, '78]

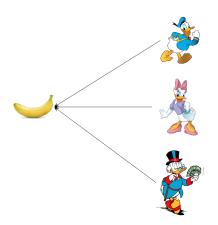
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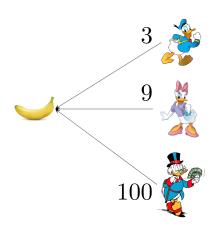
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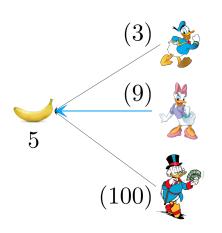
$$\mathbb{E}[\max_i X_i] = \frac{1}{\varepsilon} \cdot \varepsilon + 1 \cdot (1 - \varepsilon) = 2 - \varepsilon.$$

- ▶ Idea: Set threshold T, accept first  $X_i \ge T$ .
  - ►  $T : Pr[\max_i X_i \ge T] = \frac{1}{2}$  works [Samuel-Cahn '84].
  - ►  $T = \frac{1}{2} \cdot \mathbb{E}[\max_i X_i]$  works [Kleinberg and Weinberg '12].

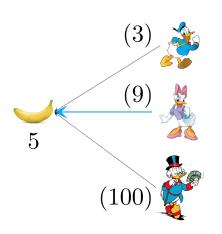








Price  $p \iff$  Threshold T in Prophet Inequality



Price  $p \iff$  Threshold T in Prophet Inequality

What about maximizing revenue? Use "virtual valuations" to design p:  $\phi(v) = v - \frac{1 - F(v)}{f(v)}$ .

[Myerson '81]

- ▶ Posted prices apply when buyers arrive *online*.
- Lots of past work on this and extensions:
  - ► [Hajiaghayi, Kleinberg and Sandholm '07]
  - ► [Chawla, Hartline, Malec and Sivan '10]
  - ► [Alaei '11]
  - ► [Babaioff, Immorlica, Lucier and Weinberg '14]
  - ▶ [Dütting, Feldman, Kesselheim and Lucier '16]
  - ► [Correa, Foncea, Pizarro and Verdugo '19]
  - ► [Correa and Cristi '23]

## **IID Prophet Inequality**

#### Fundamental questions:

- 1. What if  $\mathcal{D}_1 = \mathcal{D}_2 = \cdots = \mathcal{D}_n = \mathcal{D}$ ? Can 1/2 be improved?
- 2. Optimal (online) algorithm?
- 3. Worst-case  $\mathcal{D}$ ?

## **IID Prophet Inequality**

#### Fundamental questions:

- 1. What if  $\mathcal{D}_1 = \mathcal{D}_2 = \cdots = \mathcal{D}_n = \mathcal{D}$ ? Can 1/2 be improved?
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- 3. Worst-case  $\mathcal{D}$ ?

#### IID Prophet Inequality [Hill-Kertz '82, Correa, Foncea, Hoeksma, Oosterwijk and Vredeveld '21]]

For any  $\mathcal{D}$ ,  $\exists$  threshold stopping strategy  $\tau_1, \tau_2, \ldots, \tau_n$  that achieves  $\beta \cdot \mathbb{E}[\max_i X_i]$ , where  $\beta \approx 0.745$ , and this is tight.

Worst-case  $\mathcal{D}$ : High variance – depends on n Most of the mass is at 0 – low probability of getting a high value.

# Optimal Threshold DP

If we reach  $X_n$ , take it. Focus on  $X_{n-1}, X_n$ . What should  $\tau_{n-1}$  be?

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$$\mathbb{E}[\mathrm{OPTALG}_{n-1,n}] = (1 - F(\tau_{n-1})) \mathbb{E}[X \mid X \ge \tau_{n-1}] + F(\tau_{n-1}) \mathbb{E}[X]$$
  
 $\implies \tau_{n-1} = \mathbb{E}[X].$ 

In general, we have  $\tau_i = \mathbb{E}[OPTALG_{i+1,...,n}].$ 

#### Cost Minimization

What if objective is  $min_i X_i$ ? Same problem?

- ▶ Objective: Minimize selected value, compare against  $\mathbb{E}[\min_i X_i]$ .
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- No bounded approximation for adversarial or random order (non-IID setting)!

$$\begin{split} X_1 &= 1 \text{ w.p. } 1, & X_2 = \begin{cases} 1/\varepsilon & \text{w.p.} & \varepsilon \\ 0 & \text{w.p.} & 1-\varepsilon \end{cases} \\ & \frac{\mathbb{E}[ALG]}{\mathbb{E}[\min\{X_1,X_2\}]} = \frac{1}{\varepsilon} \end{split}$$

What about I.I.D.? Intuition: Set  $T = 2 \cdot \mathbb{E}[\min_i X_i]$ .

#### Cost Minimization

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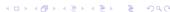
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The setting is 
$$X_1 = 1$$
 w.p.  $1$ ,  $X_2 = \begin{cases} 1/\varepsilon & \text{w.p.} & \varepsilon \\ 0 & \text{w.p.} & 1-\varepsilon \end{cases}$  
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What about I.I.D.? <u>Intuition False Intuition:</u>

Set 
$$T = 2 \cdot \mathbb{E}[\min_i X_i]$$
.

- ▶ Doesn't work!  $Pr[We are forced to select X_n] \ge c$ .
- ▶ Optimal single threshold  $T \implies \Theta$  (polylog n)-approximation. [L.-Mehta '22]



#### Is Cost Minimization hopeless?

Analyze the optimal DP. Set  $\tau_i$ , accept first  $X_i \leq \tau_i$ . Same intuition holds:  $\tau_i = \mathbb{E}[\mathrm{OPTALG}_{i+1,\dots,n}]$ . How to analyze it?

#### Is Cost Minimization hopeless?

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#### Idea

Look at "fatness" of  $\mathcal{D}$ 's tail. Captured by  $\mathcal{D}$ 's Hazard Rate.

$$h(x) = \frac{f(x)}{1 - F(x)}$$
for MAX
$$r(x) = \frac{f(x)}{F(x)}$$
for MIN

Intuition:  $h(x) = \Pr[X = x \mid X \ge x], r(x) = \Pr[X = x \mid X \le x]$  (for discrete distributions).

#### MHR Distribution

h is increasing.

Important subclass, lots of past work by economists. Good guarantees in many applications (e.g. revenue maximization in auctions).

#### Theorem [L.-Mehta '22]

In the Min-PI setting, for every distribution,

- ▶ If  $\mathbb{E}[X] = +\infty$ , the competitive ratio is infinite.
- ▶ If  $\mathbb{E}[X] < +\infty$ , there exists a constant c-approximate cost minimization prophet inequality, and we characterize the *optimal* c as the solution to a simple inequality.
- Closed form for c for special subclass of distributions.

#### Some observations:

- ightharpoonup c is distribution-dependent. Can be arbitrarily large.
- First distribution-sensitive optimal guarantees for prophet inequalities.
- Use of hazard rate in prophet inequalities as analysis tool is new.
- For MHR distributions  $\implies c = 2$ -approximation.

Why  $c = +\infty$  when  $\mathbb{E}[X] = +\infty$ ?

#### Equal-Revenue (Pareto) Distribution:

$$F(x) = 1 - 1/x$$
, with  $x \in [1, +\infty)$ .

$$\mathbb{E}[X] = 1 + \int_{1}^{\infty} (1 - F(x)) dx = +\infty$$
, but

$$\mathbb{E}[\min\{X_1, X_2\}] = 1 + \int_1^{\infty} (1 - F(x))^2 dx < +\infty.$$

▶ Due to [Lucier '22].

What if  $\mathcal D$  is independent of n? What is c if we fix  $\mathcal D$  and take  $n \to \infty$ ? What if  $\mathcal{D}$  is independent of n? What is c if we fix  $\mathcal{D}$  and take  $n \to \infty$ ?

# [Kennedy-Kertz '91]

For any  $\mathcal{D}$ , independent of n,  $\exists$  threshold stopping strategy  $\tau_1, \tau_2, \ldots, \tau_n$  that achieves  $\lambda \cdot \mathbb{E}[\max_i X_i]$ , as  $n \to \infty$ , where  $\lambda \approx 0.776$ , and this is tight.

What if  $\mathcal D$  is *independent* of n? What is c if we fix  $\mathcal D$  and take  $n \to \infty$ ?

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#### [Braun-Buttkus-Kesselheim '21]

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Let 
$$M_n = \max \{X_1, \dots, X_n\}, m_n = \min \{X_1, \dots, X_n\}.$$

Limiting distribution of  $M_n$  or  $m_n$  as  $n \to \infty$ ?

Clearly,  $\lim_{n\to\infty} M_n = +\infty$  and  $\lim_{n\to\infty} m_n = 0$ , thus we need rescaling.

## Extreme Value Theory

#### Extreme Value Theorem [Fisher-Tippett '28, Gnedenko '43]

Assume there exist sequences  $a_n > 0, b_n \in \mathbb{R}$  such that

$$\lim_{n\to\infty}F_{M_n}(a_nx+b_n)=G_{\gamma}^+(x).$$

Then,

$$G_{\gamma}^{+}(x) = \begin{cases} \exp\left(-(1+\gamma x)^{-1/\gamma}\right), & \text{if } \gamma \neq 0 \\ \exp\left(-\exp\left(-x\right)\right), & \text{if } \gamma = 0 \end{cases}$$

where 
$$\gamma = \lim_{x \to \infty} \left(\frac{1}{h(x)}\right)'$$
.

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where 
$$\gamma = \lim_{x \to \infty} \left( \frac{1}{h(x)} \right)'$$
.

- Analogue of Central Limit Theorem for MAX instead of averages.
- $\triangleright$  G: Extreme Value Distribution,  $\gamma$ : Extreme Value Index
- ► Can get similar result for MIN by  $G^-(x) = 1 G^+(-x)$ , but  $\gamma$  changes.

# IID PI via Extreme Value Theory

#### Theorem [L. '23]

Assume there exist sequences  $a_n > 0, b_n \in \mathbb{R}$  such that

$$\lim_{n\to\infty} F_{M_n}(a_n x + b_n) = G_{\gamma}^+(x) \qquad \lim_{n\to\infty} F_{m_n}(a_n x + b_n) = G_{\gamma}^-(x)$$
with  $\gamma = \lim_{x\to\infty} \left(\frac{1}{h(x)}\right)'$ .
$$\text{with } \gamma = \lim_{x\to 0^+} \left(\frac{1}{r(x)}\right)'$$
.

Then, the optimal DP achieves a competitive ratio, as  $n \to \infty$ , of

$$\min\left\{\frac{(1-\gamma)^{-\gamma}}{\Gamma(1-\gamma)},1\right\}. \qquad \qquad \left| \qquad \max\left\{\frac{(1-\gamma)^{-\gamma}}{\Gamma(1-\gamma)},1\right\}.$$

# IID PI via Extreme Value Theory

#### Theorem [L. '23]

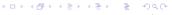
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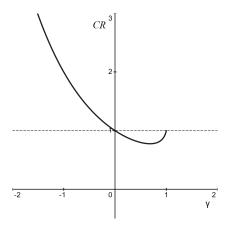
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- Unified analysis of competitive ratio for both MAX and MIN.
- ► Recovers and generalizes previous results of [Kennedy-Kertz '91] and [Braun-Buttkus-Kesselheim '21].
- Competitive ratio has same form for both MAX and MIN!



# Asymptotic Competitive Ratio



For  $\gamma \to -\infty$ , by Stirling's approximation

$$\frac{(1-\gamma)^{-\gamma}}{\Gamma(1-\gamma)} \approx e^{-\gamma}.$$

# Asymptotic Competitive Ratio

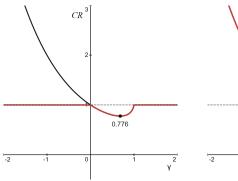


Figure: ACR( $\gamma$ ) for MAX

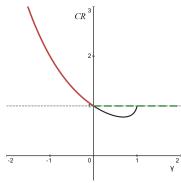


Figure: ACR( $\gamma$ ) for MIN

# Open Problems

- Extend Min-PI to multiple selection.
- Are there  $\mathcal{D}_i$  for which we can get constant approximation in the non-IID setting?
- ▶ What can you get with 1 < k < n thresholds?

## Thank You!

# Questions?

