

# On Submodular Prophet Inequalities and Correlation Gap

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September 23rd, 2021

# Prophet Inequality Problem

$$X_1 = 1, \text{ w.p. } 1$$



$$X_2 = \begin{cases} 9, & \text{w.p. } \frac{1}{9} \\ 0, & \text{w.p. } \frac{8}{9} \end{cases}$$



$$X_3 \sim U[0, 2]$$

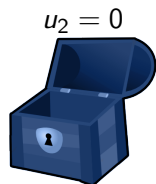


$$X_4 \sim \text{Poi}(1.5)$$



- Given independent r.v.'s  $X_1, \dots, X_n \sim D_1, \dots, D_n$ .
- $X_1, \dots, X_n$  arrive in *adversarial order*. At step  $i$ , see realization  $u_i$  of  $X_i$ .
- Decide *immediately* and *irrevocably* to select or discard  $u_i$ .
- **Goal:** Select highest possible *single value*, compare against offline optimal  $\mathbb{E}[\max_i X_i]$ .

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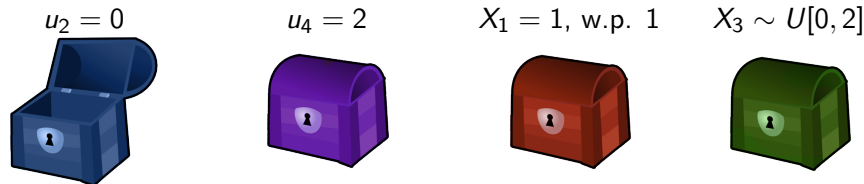


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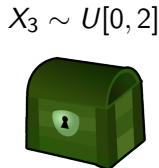
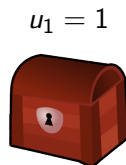
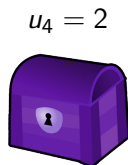
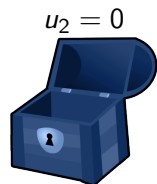
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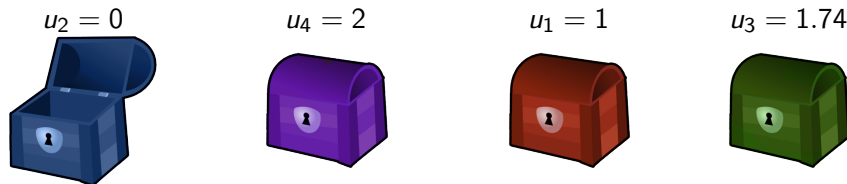
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# Prophet Inequality Problem



- Tight  $1/2$ -competitive algorithm known [KS77].
- Set threshold  $\tau = \frac{\mathbb{E}[\max_i X_i]}{2}$ , accept first value above threshold [KW12].
- Applications: Posted Prices Mechanisms [HKS07; Cha+10].
- Related to *secretary problem* (no distributional information but random order).

# Beyond Single Item ?

- Pick many elements, constraint on what you can pick.
- Additive functions:
  - Cardinality  $k$ :  $1 - \frac{1}{\sqrt{k+3}}$  [Ala14]
  - Matroid:  $\frac{1}{2}$  [KW12]
  - Matching: 0.337 [Ezr+20]
  - Downward-Closed:  $O(\log n \cdot \log r)$  [Rub16]

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  - Downward-Closed:  $O(\log n \cdot \log r)$  [Rub16]
- Several approaches. One idea: *Relaxation* + (online) *rounding*.
- Example:  $X_i = \begin{cases} v_i & \text{w.p. } p_i \\ 0 & \text{otherwise} \end{cases}$

Solve  $\max \sum_i x_i v_i$  then round  $x$   
to get feasible  $S$

$$x \in P(\mathcal{C})$$
$$\forall i \in N \quad x_i \leq p_i$$



# Beyond Additive Functions ?

- *Submodular* functions:

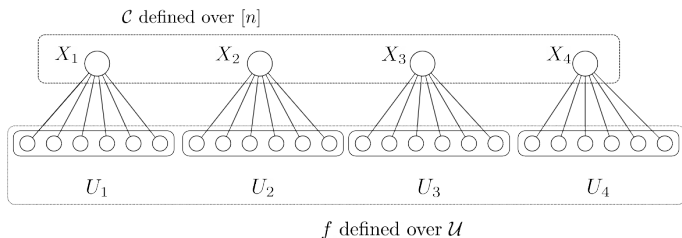
$$\forall A, B \subseteq N : f(A) + f(B) \geq f(A \cap B) + f(A \cup B).$$

- If  $A \subseteq B \implies f(A) \leq f(B)$ ,  $f$  is *monotone*.
- Assume  $f(S) \geq 0$  for all  $S$ .
- How to capture distributional information of PI model?

# Submodular Prophet Inequality

- Submodular Prophet Inequality (SPI):

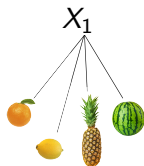
- Introduced by [RS16] to capture distributional information of  $X_i$ 's.
- $X_i$  takes one value from  $U_i$  (correlated distribution), let  $\mathcal{U} = \bigcup_{i=1}^n U_i$ .
- $f : 2^{\mathcal{U}} \rightarrow \mathbb{R}_{\geq 0}$ , submodular.
- Constraint  $\bar{\mathcal{C}}$  on  $N$  instead of  $\mathcal{U}$ .



- Model generalizes traditional prophet inequality problem.

# Submodular Prophet Inequality

- Example:



$$f(\text{orange}, \text{phone}) = 3, \quad f(\text{orange}, \text{phone}, \text{tent}) = 4$$

# Submodular Prophet Inequality

## Theorem [RS16]

$\exists O(1)$  –factor Submodular Prophet Inequalities for *single matroid* [RS16].

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Proof Idea: Given marginals  $x^*$  of OPT,

### 1 Step 1:

- “Relax”  $f$  into a continuous function, use *correlation gap*.

$$\max_{S \text{ is feasible}} f(S) \implies \max_{x \in P} F(x)$$

- Continuous Extensions:

- Multilinear Extension:  $F(x) = \mathbb{E}_{R \sim x} [f(R)]$ .

- Concave Closure:

$$f^+(x) = \max_{a_S} \left\{ \sum_S a_S f(S) \mid a_S \geq 0, \sum_S a_S = 1, \forall i, \sum_{S:i \in S} a_S = x_i \right\}.$$

- Observe  $f^+(x^*) \geq OPT$ .

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- Observe  $f^+(x^*) \geq OPT$ .

### 2 Step 2: *Round* $x^*$ to get feasible $S$ .

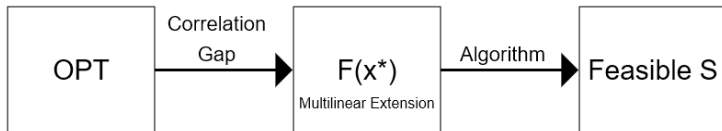
# Correlation Gap

- $\inf_{x \in [0,1]^n} \frac{F(x)}{f^+(x)}$  is *correlation gap* of  $f$ .
- For *monotone submodular*  $f$ , correlation gap is at most  $1 - \frac{1}{e}$  [Cal+11; Agr+12].
- For general submodular  $f$  (special case of subadditive), correlation gap can be arbitrarily large.
- To get around this, [RS16] define correlation gap *variant*:

$$\inf_{x \in [0,1]^n} \frac{F_{\max}(x)}{f^+(x)} \geq \frac{1}{200}, \quad \text{where } F_{\max}(x) = \max_{y \leq x} F(y).$$

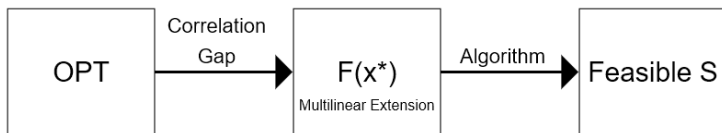
# Rounding $x^*$

- [RS16] round  $x^*$  and get bound w.r.t.  $F_{max}(x^*)$ .





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- $\exists$  rounding schemes which account for constraints: *Contention Resolution Scheme (CRS)* [CVZ11].
- *Online* CRS [FSZ16]: A CRS that rounds  $x$  in online manner.
- $\exists O(1)$ -factor OCRSs for matroids, matchings, knapsacks, their intersection, etc [FSZ16].

# Can we do better?

- 1  $\frac{1}{200}$  seems unreasonable for non-monotone correlation gap.
- 2 What about standard definition of correlation gap?
- 3 [RS16] technique only works for single matroid. What about more constraints [Luc17]?
- 4 Constant is in the thousands. Can we get improved bounds?
- 5 [RS16] assume knowledge of marginals  $x^*$  of  $OPT$ . Can we get results in polynomial time?

- 1 Via nice interpretation of known properties, we get  
$$\inf_{x \in [0,1]^n} \frac{F_{max}(x)}{f^+(x)} \geq \frac{1}{e}.$$

# Results (1/3)

- 1 Via nice interpretation of known properties, we get  $\inf_{x \in [0,1]^n} \frac{F_{\max}(x)}{f^+(x)} \geq \frac{1}{e}$ .
- 2 Parametrize by  $p = \max_i x_i$ . Use *Measured Continuous Greedy* algorithm to get

$$\frac{F(y)}{f^+(x)} \geq \begin{cases} 1 - \frac{1}{e} - p - \frac{\ln(1-p)}{e}, & p \in [0, 1 - 1/e] \\ 1/e, & p \in [1 - 1/e, 1] \end{cases}.$$

- ③ Standard definition of correlation gap:

## Theorem

For any non-negative submodular  $f : 2^N \rightarrow \mathbb{R}_{\geq 0}$  and any  $x \in [0, 1]^n$  s.t.  $\max_i x_i \leq \rho$ ,

$$\frac{F(x)}{f^+(x)} \geq (1 - \rho) \left(1 - \frac{1}{e}\right).$$

- Interesting on its own.

## ④ Rounding:

### Theorem

For any constraint, there exists an *efficient black-box reduction* from Submodular Prophet Inequality to Online Contention Resolution Schemes.

- Generalizes rounding to several constraints.
- Allows for significantly improved bounds for SPI.

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### Theorem

For any constraint, there exists an *efficient black-box reduction* from Submodular Prophet Inequality to Online Contention Resolution Schemes.

- Generalizes rounding to several constraints.
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Feasibility constraint	Competitive Ratio	
	Monotone Submodular	General Submodular
Uniform matroid of rank $k \rightarrow \infty$	1/4.3	1/17.2
Matroid	1/7.4	1/30
Matching	1/9.5	1/38
Knapsack	1/17.5	1/70
Intersection of $k$ matroids	$\Omega(1/k)$	$\Omega(1/k)$

Table 1: A summary of our results for several feasibility constraints.

- 1 *Generalize SPI framework* of [RS16] for arbitrary down-closed constraints for which we have an OCRS.
- 2 Fine-grained *correlation gap* for general non-negative submodular functions (w.r.t.  $p = \max_i x_i$ ).
- 3 Fine-grained analysis of *Measured Continuous Greedy* algorithm.
- 4 Significantly *improved bounds for SPI* for several constraints in *polynomial time*.

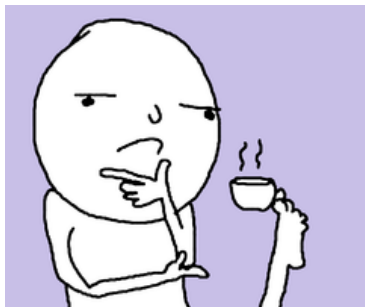


# Open Problems

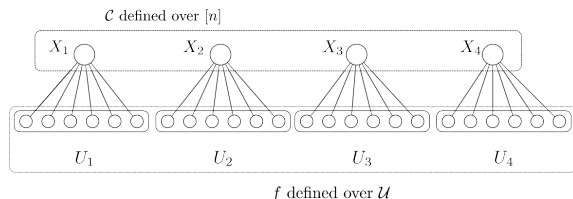
- 1 Can we get a tight  $\frac{1}{2}$ -submodular prophet inequality for a matroid constraint?
- 2 Can the  $(1 - \rho) \left(1 - \frac{1}{e}\right)$ -fine-grained correlation gap for general submodular functions be made tight?

# THANK YOU!

## QUESTIONS?



# Rounding Subtleties



- Follows framework of [RS16].
- Recall:  $\exists$  correlations! Exactly one element of  $U_i$  realizes at step  $i$ .
- **Main Idea:** Get rid of correlations by treating  $X_i \sim \text{Prod}(U_i)$ .
- Rubinfeld-Singla's approach uses O CRS on  $\mathcal{U} \implies$  only works on single matroid.
- Refined approach: Eliminate need for O CRS on  $\mathcal{U}$ , use O CRS only on  $N \implies$  works for any constraint.



Shipra Agrawal et al. “Price of Correlations in Stochastic Optimization”. In: *Operations Research* 60.1 (2012), pp. 150–162. ISSN: 0030364X, 15265463. URL: <http://www.jstor.org/stable/41476345>.



Saeed Alaei. “Bayesian Combinatorial Auctions: Expanding Single Buyer Mechanisms to Many Buyers”. In: *SIAM Journal on Computing* 43.2 (2014), pp. 930–972. DOI: 10.1137/120878422. eprint: <https://doi.org/10.1137/120878422>. URL: <https://doi.org/10.1137/120878422>.



Gruia Calinescu et al. “Maximizing a Monotone Submodular Function Subject to a Matroid Constraint”. In: *SIAM Journal on Computing* 40.6 (2011), pp. 1740–1766. DOI: 10.1137/080733991. eprint: <https://doi.org/10.1137/080733991>. URL: <https://doi.org/10.1137/080733991>.



Shuchi Chawla et al. “Multi-Parameter Mechanism Design and Sequential Posted Pricing”. In: *Proceedings of the Forty-Second ACM Symposium on Theory of Computing*. STOC '10. Cambridge, Massachusetts, USA: Association for Computing Machinery, 2010, pp. 311–320. ISBN: 9781450300506. DOI: 10.1145/1806689.1806733. URL: <https://doi.org/10.1145/1806689.1806733>.



Chandra Chekuri, Jan Vondrák, and Rico Zenklusen. “Submodular Function Maximization via the Multilinear Relaxation and Contention Resolution Schemes”. In: *Proceedings of the Forty-third Annual ACM Symposium on Theory of Computing*. STOC '11. San Jose, California, USA: ACM, 2011, pp. 783–792. ISBN: 978-1-4503-0691-1. DOI: 10.1145/1993636.1993740. URL: <http://doi.acm.org/10.1145/1993636.1993740>.



Tomer Ezra et al. “Online Stochastic Max-Weight Matching: Prophet Inequality for Vertex and Edge Arrival Models”. In: *Proceedings of the 21st ACM Conference on Economics and Computation*. EC '20. Virtual Event, Hungary: Association for Computing Machinery, 2020, pp. 769–787. ISBN: 9781450379755. DOI: 10.1145/3391403.3399513. URL: <https://doi.org/10.1145/3391403.3399513>.

## References IV



Moran Feldman, Ola Svensson, and Rico Zenklusen. “Online Contention Resolution Schemes”. In: *Proceedings of the Twenty-seventh Annual ACM-SIAM Symposium on Discrete Algorithms*. SODA '16. Arlington, Virginia: Society for Industrial and Applied Mathematics, 2016, pp. 1014–1033. ISBN: 978-1-611974-33-1. URL: <http://dl.acm.org/citation.cfm?id=2884435.2884507>.



Mohammad Taghi Hajiaghayi, Robert Kleinberg, and Tuomas Sandholm. “Automated Online Mechanism Design and Prophet Inequalities”. In: *Proceedings of the 22Nd National Conference on Artificial Intelligence - Volume 1*. AAAI'07. Vancouver, British Columbia, Canada: AAAI Press, 2007, pp. 58–65. ISBN: 978-1-57735-323-2. URL: <http://dl.acm.org/citation.cfm?id=1619645.1619656>.



Ulrich Krengel and Louis Sucheston. “Semiamarts and finite values”. In: *Bull. Amer. Math. Soc.* 83.4 (July 1977), pp. 745–747. URL: <https://projecteuclid.org:443/euclid.bams/1183538915>.



Robert Kleinberg and S. Matthew Weinberg. “Matroid Prophet Inequalities”. In: *CoRR* abs/1201.4764 (2012). arXiv: 1201.4764. URL: <http://arxiv.org/abs/1201.4764>.



Brendan Lucier. “An Economic View of Prophet Inequalities”. In: *SIGecom Exch.* 16.1 (Sept. 2017), pp. 24–47. DOI: 10.1145/3144722.3144725. URL: <https://doi.org/10.1145/3144722.3144725>.



Aviad Rubinfeld and Sahil Singla. “Combinatorial Prophet Inequalities”. In: *CoRR* abs/1611.00665 (2016). arXiv: 1611.00665. URL: <http://arxiv.org/abs/1611.00665>.





Aviad Rubinfeld. “Beyond Matroids: Secretary Problem and Prophet Inequality with General Constraints”. In: *Proceedings of the Forty-Eighth Annual ACM Symposium on Theory of Computing*. New York, NY, USA: Association for Computing Machinery, 2016, pp. 324–332. ISBN: 9781450341325. URL: <https://doi.org/10.1145/2897518.2897540>.