

On the Correlation Gap of Matroids

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Overview

- 1 What is Correlation Gap (\mathcal{CG})?
- 2 \mathcal{CG} of Weighted Matroid Rank
 - ▶ State of the art
 - ▶ Role of \mathcal{CG} in approximation algorithms
- 3 Main Results
 - ▶ \mathcal{CG} of weighted matroid rank \geq \mathcal{CG} of matroid rank
 - ▶ Improved lower bound for \mathcal{CG} of matroid rank
- 4 Conclusion and Future Directions

Extensions of a Submodular Function

- Let $f : 2^E \rightarrow \mathbb{R}$ be a **monotone submodular** function.
- An **extension** of f is a **continuous** function $h : [0, 1]^E \rightarrow \mathbb{R}$ such that for every $x \in [0, 1]^E$,

$$h(x) = \mathbb{E}_\lambda[f(S)]$$

where λ is a probability distribution over 2^E with marginals x .

Multilinear Extension: Sample $i \in E$ with probability x_i independently

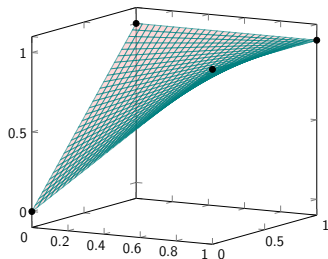
$$F(x) = \sum_{S \subseteq E} f(S) \prod_{i \in S} x_i \prod_{i \notin S} (1 - x_i).$$

Concave Extension: Distribution which maximizes expectation

$$\hat{f}(x) = \max_{\lambda} \left\{ \sum_{S \subseteq E} \lambda_S f(S) : \sum_{S \ni i} \lambda_S = x_i \forall i \in E, \sum_{S \subseteq E} \lambda_S = 1, \lambda \geq 0 \right\}.$$

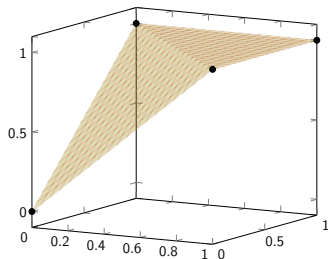
Example

- Let f be the rank function of a rank-1 uniform matroid on 2 elements.



$F(x)$

- ▶ Monotone.
- ▶ Concave along any e_j .
- ▶ Convex along any $e_i - e_j$.



$\hat{f}(x)$

- ▶ Monotone.
- ▶ Concave.
- ▶ Piecewise affine.

Correlation Gap

Def: [Agrawal et al. '12] The **correlation gap** of f is the ratio

$$CG(f) = \min_{x \in [0,1]^E} \frac{F(x)}{\hat{f}(x)}.$$

- [Calinescu et al. '07] For any **monotone submodular** function f ,

$$CG(f) \geq 1 - \frac{1}{e}.$$

- Tight for the rank function of a rank-1 uniform matroid. Letting $n = |E|$,

$$\frac{F\left(\frac{1}{n}\mathbb{1}\right)}{\hat{f}\left(\frac{1}{n}\mathbb{1}\right)} = \frac{1 - \left(1 - \frac{1}{n}\right)^n}{n \cdot \frac{1}{n}} \rightarrow 1 - \frac{1}{e} \quad \text{as } n \rightarrow \infty.$$

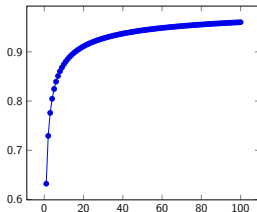
CG of Weighted Matroid Rank

Def: Given a matroid $\mathcal{M} = (E, \mathcal{I})$ and weights $w \in \mathbb{R}_+^E$, the **weighted matroid rank** function is

$$r_w(S) := \max\{w(T) : T \subseteq S, T \in \mathcal{I}\}.$$

- r_w is monotone submodular $\implies \text{CG}(r_w) \geq 1 - e^{-1}$.
- [Yan '11, Barman et al. '20] For a rank- ℓ **uniform matroid**, its (unweighted) rank function r has correlation gap

$$\text{CG}(r) = 1 - \frac{e^{-\ell} \ell^\ell}{\ell!}.$$



\mathcal{CG} is Approximation Guarantee

Problem: [Calinescu et al. '07] Given a sum of weighted matroid rank functions $f = \sum_{i=1}^m f_i$ and a matroid $\mathcal{M} = (E, \mathcal{I})$, solve

$$\max_{S \in \mathcal{I}} f(S).$$

- NP-hard as it captures the **maximum coverage** problem:

MaxCover: Given a family of sets $E \subseteq 2^N$, select k sets to cover as many elements in N as possible.

Reduction: Let $\mathcal{M} = (E, \mathcal{I})$ be a uniform matroid of rank k . For each $i \in N$, define $f_i : 2^E \rightarrow \mathbb{Z}_+$ as

$$f_i(S) = \begin{cases} 1, & \text{if } \exists \text{ a set in } S \text{ containing } i, \\ 0, & \text{otherwise.} \end{cases}$$

- Each f_i is the weighted rank function of a rank-1 uniform matroid.

CG is Approximation Guarantee

- A generalization of **MaxCover** is the **maximum multicoverage** problem.

Max- ℓ -Cover: Given a family of sets $E \subseteq 2^N$, select k sets to cover as many elements in N as possible, where each element is counted **up to ℓ times**.

Reduction: Let $\mathcal{M} = (E, \mathcal{I})$ be a uniform matroid of rank k . For each $i \in N$, define $f_i : 2^E \rightarrow \mathbb{Z}_+$ as

$$f_i(S) = \min(\# \text{ of sets in } S \text{ containing } i, \ell).$$

- Each f_i is the weighted rank function of a rank- ℓ uniform matroid.
- [Barman et al. '20] gave a tight $1 - \frac{e^{-\ell} \ell^\ell}{\ell!}$ approximation algorithm.
- Applications in list decoding and approval voting.

\mathcal{CG} is Approximation Guarantee

Algorithm: [Calinescu et al. '07, Shioura '09]

- 1 Let x^* be an optimal solution to the following LP

$$\max \left\{ \sum_{i=1}^m \hat{f}_i(x) : x \in \text{independent set polytope of } \mathcal{M} \right\}.$$

- 2 Round x^* to $x' \in \{0, 1\}^E$ using **pipage rounding**.

Analysis:

$$f(x') = F(x') \geq F(x^*) = \sum_{i=1}^m F_i(x^*) \geq \sum_{i=1}^m \mathcal{CG}(f_i) \hat{f}_i(x^*) \geq \min_{i \in [m]} \mathcal{CG}(f_i) \cdot \text{OPT}.$$

- \mathcal{CG} is also the **approximation guarantee** in **sequential posted price mechanism** and **contention resolution schemes**.

Main Results

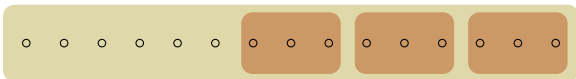
Theorem 1

For any matroid $\mathcal{M} = (E, \mathcal{I})$ with rank function r ,

$$\inf_{w \in \mathbb{R}_+^E} \mathcal{CG}(r_w) = \mathcal{CG}(r).$$

Goal: Identify matroid parameters which govern $\mathcal{CG}(r)$.

- Parameterization by **rank** ρ or **girth** γ only yields $\mathcal{CG}(r) = 1 - 1/e$.
 - ▶ Partition matroid with ρ rank-1 parts.
 - ▶ Union of a **rank**-($\gamma - 1$) **uniform matroid** and **many rank-1 uniform matroids**.

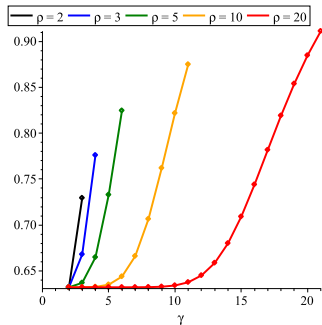
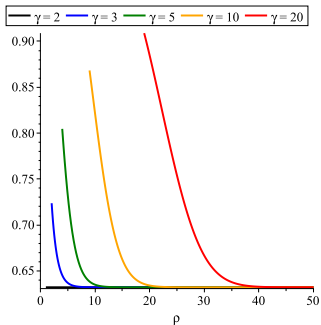


Main Results

Theorem 2

For any matroid \mathcal{M} with rank function r , rank ρ , and girth $\gamma > 1$,

$$CG(r) \geq 1 - \frac{1}{e} + \frac{e^{-\rho}}{\rho} \left(\sum_{i=0}^{\gamma-2} (\gamma - 1 - i) \left[\binom{\rho}{i} (e-1)^i - \frac{\rho^i}{i!} \right] \right) \geq 1 - \frac{1}{e}.$$



Proof Ideas

Theorem 1:

- Let $w \in \mathbb{R}_+^E$ where $\mathcal{CG}(r_w) < \mathcal{CG}(r)$ with the **fewest** distinct values.
- Can increase weights in a value class without increasing $\mathcal{CG}(r_w)$. \Downarrow

Theorem 2:

- Let r be the rank function of a matroid \mathcal{M} with rank ρ and girth γ .
- \exists a point x in the independent set polytope of \mathcal{M} such that

$$\mathcal{CG}(r) = \frac{R(x)}{\hat{r}(x)} = \frac{R(x)}{\mathbb{1}^\top x}$$

- To bound $R(x)$, decompose $r = g + h$ such that g is the rank function of a **rank- ℓ uniform matroid**, where $\ell := \gamma - 1$.
 - ▶ Bounding $G(x)$: simple application of [Yan '11, Barman et al. '20]
 - ▶ Bounding $H(x)$: Poisson clock analysis [Calinescu et al. '07]

Poisson Clock Analysis

- Construct a random set $Q(1)$ as follows:
 - ▶ Put a **Poisson clock** of rate x_i on each element i (independent).
 - ▶ From time $t = 0$ to 1, add i to $Q(t)$ whenever its clock rings.



- By monotonicity of h , $H(x) \geq H(1 - e^{-x}) = \mathbb{E}[h(Q(1))]$.
- In [Calinescu et al. '07], $\frac{d}{dt}\mathbb{E}[h(Q(t))] \geq \hat{h}(x) - \mathbb{E}[h(Q(t))]$ for all t .

$$\Rightarrow \mathbb{E}[h(Q(1))] \geq (1 - e^{-1})\hat{h}(x).$$

- For us, $\frac{d}{dt}\mathbb{E}[h(Q(t))] \geq \hat{r}(x) - \ell - \mathbb{E}[h(Q(t))]$ only if $|Q(t)| \geq \ell$.

$$\Rightarrow \mathbb{E}[h(Q(1)) \mid |Q(t)| \geq \ell \forall t \geq T'] \geq (1 - e^{T'-1})(\hat{r}(x) - \ell).$$

Conclusion

Summary of Results:

- 1 \mathcal{CG} of weighted matroid rank is minimized under **uniform weights**.
- 2 Improved lower bound on $\mathcal{CG}(r)$ parameterized by **rank** and **girth**.
 - ▶ When girth is fixed, $\mathcal{CG} \downarrow$ as rank \uparrow .
 - ▶ When rank is fixed, $\mathcal{CG} \uparrow$ as girth \uparrow .

Future Directions:

- 1 Upper and lower bounds do not match. There exists a rank- ρ girth- γ matroid with

$$\mathcal{CG}(r) \leq 1 - \frac{1}{e} + \frac{\gamma-1}{e\rho}.$$

- 2 Find better matroid parameters that govern $\mathcal{CG}(r)$.