

A cross-intersection theorem for vector spaces based on semidefinite programming

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The q -Erdős–Ko–Rado theorem

- V : a vector space/ \mathbb{F}_q with $\dim V = n$
- $\binom{V}{k} := \{x \leq V : \dim x = k\}$
- $\mathcal{F} \subseteq \binom{V}{k}$: t -intersecting $\stackrel{\text{def}}{\iff} \dim x \cap y \geq t \ (\forall x, y \in \mathcal{F})$

Theorem (Hsieh, 1975; T., 2006)

- $\mathcal{F} \subseteq \binom{V}{k}$: t -intersecting, where $n \geq 2k$
- Then

$$|\mathcal{F}| \leq \binom{n-t}{k-t}. \quad (\#)$$

- Equality holds in (#) \iff One of the following holds:
 - 1 $\exists z \in \binom{V}{t}$ s.t. $\mathcal{F} = \{x \in \binom{V}{k} : z \subseteq x\}$,
 - 2 $n = 2k$, and $\exists z \in \binom{V}{2k-t}$ s.t. $\mathcal{F} = \{x \in \binom{V}{k} : x \subseteq z\}$.

The q -Erdős–Ko–Rado theorem

Remark

- The bound (#) (for all n, k, q, t) is due to Frankl and Wilson (1986).
- For $t = 1$, the q -EKR theorem was proved independently by Godsil and Newman (2006).

- $\mathcal{F} \subseteq \binom{V}{k}, \mathcal{G} \subseteq \binom{V}{\ell}$: **cross-(1-)intersecting**
 $\stackrel{\text{def}}{\iff} x \cap y \neq \emptyset \ (\forall x \in \mathcal{F}, \forall y \in \mathcal{G})$

Theorem (Suda–T., 2013)

- $\mathcal{F} \subseteq \binom{V}{k}, \mathcal{G} \subseteq \binom{V}{\ell}$: cross-intersecting, where $n \geq 2k, 2\ell$
- Then

$$|\mathcal{F}| |\mathcal{G}| \leq \binom{n-1}{k-1} \binom{n-1}{\ell-1}. \quad (\text{b})$$

- Equality holds in (b) \iff One of the following holds:
 - 1 $\exists z \in \binom{V}{1}$ s.t. $\mathcal{F} = \{x \in \binom{V}{k} : z \subseteq x\}, \mathcal{G} = \{x \in \binom{V}{\ell} : z \subseteq x\},$
 - 2 $n = 2k = 2\ell,$ and $\exists z \in \binom{V}{2k-1}$ s.t. $\mathcal{F} = \mathcal{G} = \{x \in \binom{V}{k} : x \subseteq z\}.$

- This is a q -analogue of a result of Pyber (1986) and Matsumoto and Tokushige (1989), the proof of which uses the **Kruskal–Katona theorem**.
- Our proof is algebraic in nature and uses the duality of semidefinite programming.

How to prove the q -EKR theorem (Part I)

For simplicity, we assume $t = 1$.

- $\Gamma_k = qK_{n:k}$: the q -Kneser graph
 - $V(\Gamma_k) = \binom{V}{k}$
 - $E(\Gamma_k) = \{(x, y) : x, y \in \binom{V}{k}, x \cap y = \emptyset\}$
- $A_k \in \mathbb{R}^{\binom{V}{k} \times \binom{V}{k}}$: the adjacency matrix of Γ_k :

$$(A_k)_{xy} := \begin{cases} 1 & \text{if } x \sim y \\ 0 & \text{otherwise} \end{cases} \quad (x, y \in \binom{V}{k})$$

- $I_k \in \mathbb{R}^{\binom{V}{k} \times \binom{V}{k}}$: the identity matrix
- $J_k \in \mathbb{R}^{\binom{V}{k} \times \binom{V}{k}}$: the all 1's matrix
- $Y \bullet Z := \text{trace}(Y^T Z)$ ($\forall Y, Z \in \mathbb{R}^{\binom{V}{k} \times \binom{V}{k}}$)

How to prove the q -EKR theorem (Part I)

- $\mathcal{F} \subseteq \binom{[V]}{k}$: a (1-)intersecting family
 \iff an independent set of Γ_k
- $\varphi \in \mathbb{R}^{\binom{[V]}{k}}$: the (column) characteristic vector of \mathcal{F} :

$$\varphi_x = \begin{cases} 1 & \text{if } x \in \mathcal{F} \\ 0 & \text{otherwise} \end{cases} \quad (x \in \binom{[V]}{k})$$

- $X := \frac{1}{\|\varphi\|^2} \varphi \varphi^\top \in \mathbb{R}^{\binom{[V]}{k} \times \binom{[V]}{k}}$: nonnegative & positive semidefinite
- $X \bullet I_k = 1$, $X \bullet J_k = |\mathcal{F}|$
- $X \bullet A_k = \frac{1}{\|\varphi\|^2} \varphi^\top A_k \varphi = 0$

How to prove the q -EKR theorem (Part I)

- Consider the following SDP problem:

$$\vartheta_k = \max_X X \bullet J_k$$

subject to

- $X \bullet I_k = 1,$
 - $X \bullet A_k = 0,$
 - X : nonnegative & positive semidefinite.
- Then $|\mathcal{F}| \leq \vartheta_k.$

Remark

ϑ_k = the strengthening of Lovász's ϑ -function bound due to Schrijver (1979).

How to prove our theorem (Part I)

(Caution: If $k = \ell$ then we view $[V_k]$ and $[V_\ell]$ as distinct copies.)

- $\Gamma_{k,\ell}$: a bipartite graph
 - $V(\Gamma_{k,\ell}) = [V_k] \cup [V_\ell]$
 - $E(\Gamma_{k,\ell}) = \{(x, y), (y, x) : x \in [V_k], y \in [V_\ell], x \cap y = 0\}$
- $A_{k,\ell} \in \mathbb{R}([V_k] \cup [V_\ell]) \times ([V_k] \cup [V_\ell])$: the adjacency matrix of $\Gamma_{k,\ell}$:

$$A_{k,\ell} = \begin{pmatrix} 0_k & * \\ * & 0_\ell \end{pmatrix}$$

- $J_{k,\ell} \in \mathbb{R}([V_k] \cup [V_\ell]) \times ([V_k] \cup [V_\ell])$: the adjacency matrix of the complete bipartite graph with bipartition $[V_k] \cup [V_\ell]$:

$$J_{k,\ell} = \begin{pmatrix} 0_k & J \\ J & 0_\ell \end{pmatrix}$$

How to prove our theorem (Part I)

- $\mathcal{F} \subseteq [V_k]$, $\mathcal{G} \subseteq [V_\ell]$: cross-intersecting families
 $\iff \mathcal{F} \cup \mathcal{G}$: an independent set of $\Gamma_{k,\ell}$
- $\varphi \in \mathbb{R}^{[V_k]}$: the characteristic vector of \mathcal{F}
- $\psi \in \mathbb{R}^{[V_\ell]}$: the characteristic vector of \mathcal{G}
- $X := \begin{pmatrix} \frac{1}{\|\varphi\|^2} \varphi \varphi^\top & \frac{1}{\|\varphi\| \|\psi\|} \varphi \psi^\top \\ \frac{1}{\|\varphi\| \|\psi\|} \psi \varphi^\top & \frac{1}{\|\psi\|^2} \psi \psi^\top \end{pmatrix} \in \mathbb{R}^{([V_k] \cup [V_\ell]) \times ([V_k] \cup [V_\ell])}$:
nonnegative & positive semidefinite
- $X \bullet I_k = X \bullet I_\ell = 1$, $\frac{1}{2} X \bullet J_{k,\ell} = \sqrt{|\mathcal{F}| |\mathcal{G}|}$
- $X \bullet A_{k,\ell} = 0$

How to prove our theorem (Part I)

- Consider the following SDP problem:

$$\vartheta_{k,\ell} = \frac{1}{2} \max_X X \bullet J_{k,\ell}$$

subject to

- 1 $X \bullet I_k = X \bullet I_\ell = 1,$
 - 2 $X \bullet A_{k,\ell} = 0,$
 - 3 X : nonnegative & positive semidefinite.
- Then $|\mathcal{F}| |\mathcal{G}| \leq (\vartheta_{k,\ell})^2.$

$\vartheta_{k,\ell}$ is a “bipartite variant” of Lovász’s ϑ -function bound !!

How to prove the q -EKR theorem (Part II)

- We prove the EKR bound $|\mathcal{F}| \leq \binom{n-1}{k-1}$ by constructing an optimal feasible solution to the dual program of ϑ_k .
- To this end, we notice that

$$I_k, J_k, A_k \in \mathcal{A}$$

where \mathcal{A} = the **Bose–Mesner algebra** of the Grassmann graph $J_q(n, k)$ (= the commutant of $GL(V)$ acting on $\begin{bmatrix} V \\ k \end{bmatrix}$)

- By projecting the variable X to \mathcal{A} , we may assume

$$X \in \mathcal{A}.$$

- Since \mathcal{A} is a **commutative** matrix $*$ -algebra, it is simultaneously diagonalized.
 $\Rightarrow \vartheta_k$ turns to an LP !!

How to prove the q -EKR theorem (Part II)

- The dual program is given by

$$\vartheta_k = \min_{\alpha, \gamma, Z} \alpha$$

subject to

- $\alpha I_k - \gamma A_k - J_k - Z$: positive semidefinite,
 - Z : nonnegative.
- After the reduction, it can be shown that there is a **unique optimal feasible solution** to the dual program:

$$\alpha = \begin{bmatrix} n-1 \\ k-1 \end{bmatrix}, \quad \gamma = -q^{-k^2+k} \cdot \frac{\begin{bmatrix} n-1 \\ k-1 \end{bmatrix}}{\begin{bmatrix} n-k-1 \\ k-1 \end{bmatrix}}, \quad Z = 0.$$

(More on this later.)

How to prove our theorem (Part II)

- We prove the bound $|\mathcal{F}||\mathcal{G}| \leq \binom{n-1}{k-1} \binom{n-1}{\ell-1}$ by constructing an optimal feasible solution to the dual program of $\vartheta_{k,\ell}$.
- To this end, we notice that

$$I_k, I_\ell, J_{k,\ell}, A_{k,\ell} \in \mathcal{C}$$

where $\mathcal{C} =$ the **coherent algebra** of the commutant of $GL(V)$ acting on $\begin{bmatrix} V \\ k \end{bmatrix} \cup \begin{bmatrix} V \\ \ell \end{bmatrix}$

- By projecting the variable X to \mathcal{C} , we may assume

$$X \in \mathcal{C}.$$

- \mathcal{C} is a matrix $*$ -algebra, and it is simultaneously **block-diagonalized** into (at most) 2×2 matrices.
 $\Rightarrow \vartheta_{k,\ell}$ turns to a drastically smaller SDP !!

How to prove our theorem (Part II)

- The dual program is given by

$$\vartheta_{k,\ell} = \min_{\alpha, \beta, \gamma, Z} \alpha + \beta$$

subject to

- 1 $\alpha I_k + \beta I_\ell - \gamma A_{k,\ell} - \frac{1}{2} J_{k,\ell} - Z$: positive semidefinite,
 - 2 Z : nonnegative.
- After the reduction, it can be shown that there is a **unique one-parameter family** of optimal feasible solutions to the dual program:

$$\alpha = \beta = \frac{1}{2} \begin{bmatrix} n-1 \\ k-1 \end{bmatrix}^{\frac{1}{2}} \begin{bmatrix} n-1 \\ \ell-1 \end{bmatrix}^{\frac{1}{2}}, \quad \gamma = b(\lambda), \quad Z = a(\lambda)A_k + \lambda A_\ell,$$

where we are assuming $k \geq \ell$, and ... (!)

How to prove our theorem (Part II)

... the functions $a(\lambda)$ and $b(\lambda)$ are given by

$$\begin{aligned}q^{k^2}(q^k - 1) \begin{bmatrix} n-k \\ k \end{bmatrix} \cdot a(\lambda) &= \frac{1}{2}q^\ell(q^{k-\ell} - 1) \begin{bmatrix} n-1 \\ k-1 \end{bmatrix}^{\frac{1}{2}} \begin{bmatrix} n-1 \\ \ell-1 \end{bmatrix}^{\frac{1}{2}} \\ &\quad + q^{\ell^2}(q^\ell - 1) \begin{bmatrix} n-\ell \\ \ell \end{bmatrix} \lambda, \\ q^{k\ell} \begin{bmatrix} n-k \\ \ell \end{bmatrix} \cdot b(\lambda) &= -\frac{1}{2}q^\ell \begin{bmatrix} n-1 \\ \ell \end{bmatrix} - q^{\ell^2} \begin{bmatrix} n-\ell \\ \ell \end{bmatrix} \begin{bmatrix} n-1 \\ \ell-1 \end{bmatrix}^{\frac{1}{2}} \begin{bmatrix} n-1 \\ k-1 \end{bmatrix}^{-\frac{1}{2}} \lambda\end{aligned}$$

for sufficiently small $\lambda \geq 0$.

Compare this with the unique solution of ϑ_k !!

- In both of the theorems, the optimal feasible solutions to the dual programs, together with the duality of LP / SDP, provide **enough information** about the characteristic vectors of optimal (cross-)intersecting families.
- By “enough information” I do NOT mean that the characterization of optimal families is easy.

[Recall that the q -EKR theorem was proved in full generality only in 2006.]

- Indeed, the discussions in this part are very typical in **Delsarte theory**.

Delsarte theory (1973)

- studies in a unified manner various combinatorial objects (e.g., codes, designs) whose underlying spaces have “strong” symmetry / regularity.
- bounds the value of a numerical parameter (e.g., size, index) of such objects.
- shows that optimal (or nearly optimal) objects satisfy certain additional regularity.
- then in some cases classifies the optimal (or nearly optimal) objects.

Example

- Objects : intersecting families \mathcal{F}
- Underlying space : $\begin{bmatrix} V \\ k \end{bmatrix} \curvearrowright GL(V)$

“Classical” Delsarte theory

- Commutative matrix $*$ -algebras (Bose–Mesner algebras)
- Linear programming

Example

- LP bound on the size of a code (Delsarte, 1972)
 - sphere-packing bound
 - Singleton bound
 - Plotkin bound
 - McEliece–Rodemich–Rumsey–Welch bound
- Lloyd’s theorem (Delsarte, 1973)
- Erdős–Ko–Rado theorem (Wilson, 1984)

“Quantum” Delsarte theory (still in its infancy)

- Noncommutative matrix $*$ -algebras (e.g., **coherent algebras**, **Terwilliger algebras**)
- Semidefinite programming

Example

- SDP bound on the size of a code (Schrijver, 2005; Gijswijt–Schrijver–T., 2006)
- Today’s theorem

EKR theorems for distance-regular graphs

- $J_q(n, k)$: the **Grassmann graph**:
 - $V(J_q(n, k)) = \begin{bmatrix} V \\ k \end{bmatrix}$
 - $E(J_q(n, k)) = \{(x, y) : x, y \in \begin{bmatrix} V \\ k \end{bmatrix}, \dim x \cap y = k - 1\}$
- It follows that
$$x, y : \text{at distance } i \iff \dim x \cap y = k - i$$
- Hence $\Gamma_k = qK_{n:k}$: the **distance- k graph** of $J_q(n, k)$

$J_q(n, k)$ is an example of a **Q -polynomial distance-regular graph**:
“very strong regularity” + “very nice structure of the eigenspaces”

EKR theorems for distance-regular graphs

- Every Q -polynomial DRG Γ (with diameter k) is associated with the **parameter array**:

$$(\{\theta_i\}_{i=0}^k; \{\theta_i^*\}_{i=0}^k; \{\varphi_i\}_{i=1}^k; \{\phi_i\}_{i=1}^k).$$

[The θ_i are the distinct eigenvalues of Γ .]

Example

For $J_q(n, k)$, the parameter array is of the **dual q -Hahn type**:

- $\theta_i = \theta_0 + h(1 - q^i)(1 - sq^{i+1})q^{-i} \quad (0 \leq i \leq k)$
- $\theta_i^* = \theta_0^* + h^*(1 - q^i)q^{-i} \quad (0 \leq i \leq k)$
- $\varphi_i = hh^*q^{1-2i}(1 - q^i)(1 - q^{i-k-1})(1 - rq^i) \quad (1 \leq i \leq k)$
- $\phi_i = hh^*q^{k+2-2i}(1 - q^i)(1 - q^{i-k-1})(s - rq^{i-k-1}) \quad (1 \leq i \leq k)$

EKR theorems for distance-regular graphs

- The SDP problem ϑ_k can be defined for t -intersecting families ($t \geq 2$) as well, and for any Q -polynomial DRGs.
- After the reduction to LP:
 - “ X : positive semidefinite”
 - $\rightarrow k + 1$ nonnegativity constraints
(indexed $0, 1, \dots, k$)
- $f_j =$ the j^{th} component of the unique optimal feasible solution to the dual LP ($0 \leq j \leq k$)

EKR theorems for distance-regular graphs

Theorem (T., 2012)

$f_0 = 1, f_1 = \dots = f_t = 0$, and

$$f_j = \frac{\eta_{k-t}(\theta_0)}{\eta_k(\theta_0)\eta_t^*(\theta_0^*)} \frac{\phi_{k-j+1} \dots \phi_k}{\varphi_2 \dots \varphi_j(\theta_j - \theta_0)} \left(\sum_{\ell=t+1}^j \frac{\tau_\ell(\theta_j)\eta_{\ell-1}^*(\theta_0^*)\vartheta_\ell}{\phi_{k-\ell+1} \dots \phi_{k-t}} \right)$$

for $t+1 \leq j \leq k$, where

$$\begin{aligned} \tau_i(z) &= (z - \theta_0) \dots (z - \theta_{i-1}), & \eta_i(z) &= (z - \theta_d) \dots (z - \theta_{d-i+1}), \\ \tau_i^*(z) &= (z - \theta_0^*) \dots (z - \theta_{i-1}^*), & \eta_i^*(z) &= (z - \theta_d^*) \dots (z - \theta_{d-i+1}^*), \end{aligned}$$

$$\vartheta_i = \sum_{h=0}^{i-1} \frac{\theta_h - \theta_{d-h}}{\theta_0 - \theta_d}.$$

Remark

- The f_j ($t + 1 \leq j \leq k$) are expressed as ${}_4\phi_3$ basic hypergeometric series (including their special / limiting cases).
- Using this result, the EKR theorem can be proved **in a unified manner** for several families of Q -polynomial DRGs (T., 2012), e.g.,
 - Johnson graphs (Wilson, 1984) \leftrightarrow original EKR
 - Grassmann graphs (Hsieh, 1975; T., 2006) \leftrightarrow q -EKR
 - Hamming graphs (Moon, 1982) \leftrightarrow integer sequences
 - bilinear forms graphs (Huang, 1987; T., 2006)
 - twisted Grassmann graphs (T., 2012)

Summary & future work

	EKR ($t = 1$)	cross-intersection
constraints	only 1×1 matrices (i.e., LP)	involve 2×2 matrices
optimal solutions to dual program	unique	1-parameter family

Summary & future work

	EKR ($t \geq 2$)	cross t -intersection
constraints	only 1×1 matrices (i.e., LP)	involve 2×2 matrices
optimal solutions to dual program	unique	t -parameter family

The description of the above t -parameter family involves 3 parameter arrays, and it is too complicated to be stated here.