



Euler Graphs as Basis for Higher-Order Derivatives of Spectral Functions

Sebastian Krämer

Definition

A *spectral function* maps a matrix to some *symmetric* function of its eigenvalues,

$$F = f \circ \text{eig} : M \subseteq \mathbb{C}^{n \times n} \rightarrow \mathbb{C}, \quad f : \text{eig}(M) \subseteq \mathbb{C}^n \rightarrow \mathbb{C} \quad (\text{sym}),$$

where eig maps to the roots with multiplicity of the characteristic polynomial of X .

Spectral functions are conjugation invariant:

$$F(T\Lambda T^{-1}) = F(\Lambda), \quad T \in \text{GL}(n).$$

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Simplest class are trace functions,

$$F_g(X) = \text{trace } g(X) = \sum_{i \in [n]} g(\lambda_i), \quad f_g(x) = \sum_{i \in [n]} g(x_i), \quad g : D \subseteq \mathbb{C} \rightarrow \mathbb{C}.$$

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Generally,

$$f(x) = f_{\sim} \left(\underbrace{[x]}_{\text{multiset of roots in } \mathbb{C}^n / S_n} \right) = F(\text{diag}(x)).$$

Spectral Functions — About Symmetry

For each X with distinct $\lambda := \text{eig}(X)$, there is an analytic $\text{eig}^\circ : U_X \rightarrow U_\lambda \subset \mathbb{C}^n$,

$$f = f^\circ \circ \text{sort}_{U_\lambda}, \quad \text{eig}^\circ = \text{sort}_{U_\lambda} \circ \text{eig}, \quad \text{sort}_{U_\lambda} : \left(\bigcup_{i \in [n]} U_{\lambda_i} \right)^n \rightarrow U_\lambda$$

such that

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where $f^\circ : U_\lambda \rightarrow \mathbb{C}$ is *symmetric* (with nothing to permute).

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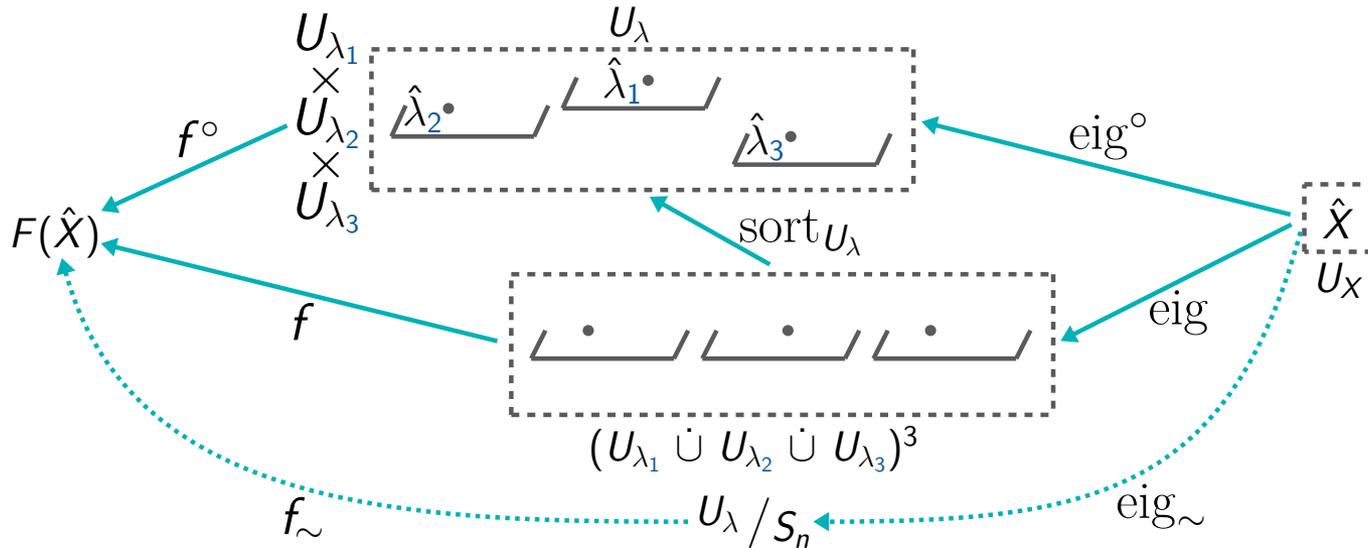
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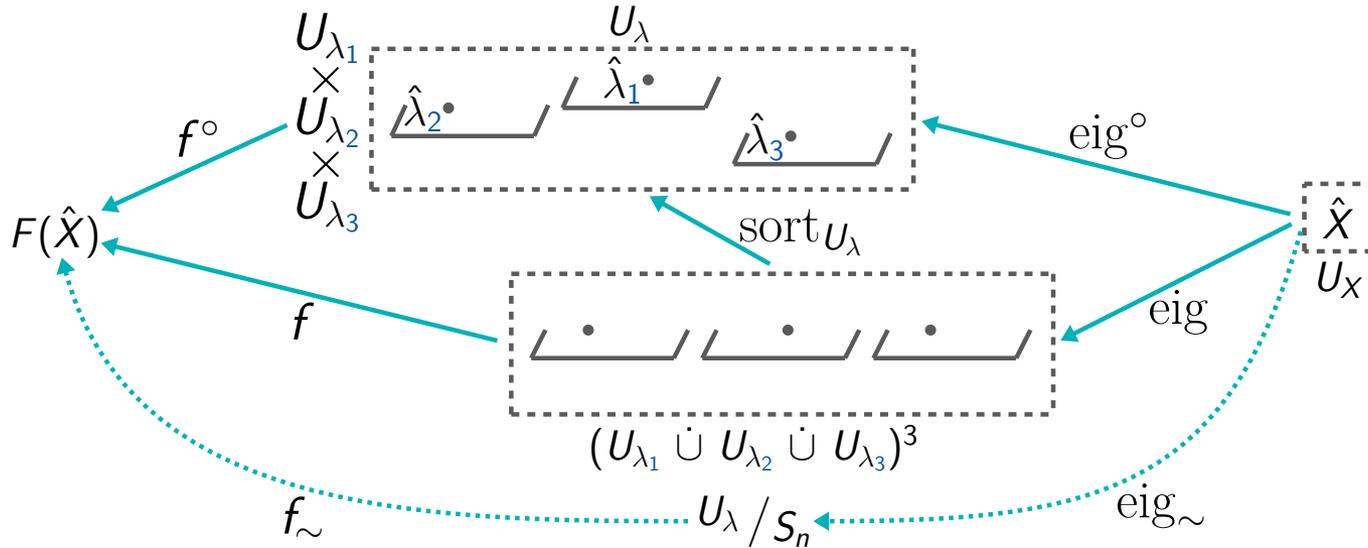
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Spectral Functions — About Symmetry



More general, for

$$\mathcal{D} = D_1 \times \dots \times D_n \subseteq \mathbb{C}^n, \quad D_i \cap D_j := \emptyset \vee D_i = D_j$$

define

$$\mathcal{M}_{\mathcal{D}} := \{ T \text{diag}(\lambda) T^{-1} \mid T \in GL(n), \lambda \in \mathcal{D} \}$$

and let

$$f^\circ : \mathcal{D} \rightarrow \mathbb{C} \text{ (domain dependent sym)}, \quad F : \mathcal{M}_{\mathcal{D}} \rightarrow \mathbb{C}.$$

Theorem (basically Riemann's Removable Singularity Theorem)

$f^\circ : \mathcal{D} \rightarrow \mathbb{C}$ analytic and *d.d.sym* $\Rightarrow F = f^\circ \circ \text{eig}^\circ : \mathcal{M}_{\mathcal{D}} \rightarrow \mathbb{C}$ analytic

Proof:

- ▶ F is continuous on all of $\mathcal{M}_{\mathcal{D}}$ as f° and eig° are continuous.
- ▶ Locally, around distinct eigenvalues, eig° and f° are analytic.
- ▶ All other points are the root set of a polynomial in entries of X .
- ▶ By Riemann's theorem, F is analytic on the whole of $\mathcal{M}_{\mathcal{D}}$.

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Example

For $\lambda_1 \in D_1$ with $D_1 \cap D_{\neq 1} = \emptyset$,

$$f^\circ : \mathcal{D} = D_1 \times D_{\neq 1} \times \dots \times D_{\neq 1} \rightarrow \mathbb{C}, \quad f^\circ(\lambda) := \lambda_1$$

is an analytic (and d.d.sym) trace function with

$$f^\circ(\lambda) = g(\lambda_1) + \dots + g(\lambda_n), \quad g : D_1 \cup D_{\neq 1} \rightarrow \mathbb{C}, \quad g(x) = \chi_{D_1}(x)x$$

Spectral functions on symmetric matrices:

Theorem ([J.Sylvester '85, M.Šilhavý '00])

$f^\circ : \mathcal{D} \subset \mathbb{R}^n \rightarrow \mathbb{R}$ in C^k and *d.d.sym* $\Rightarrow F = f^\circ \circ \text{eig}^\circ : \mathcal{M}_{\mathcal{D}, \text{sym}} \rightarrow \mathbb{R}$ in C^k .

Proofs (for $\mathcal{M}_{\mathcal{D}, \text{sym}} = \mathbb{R}_{\text{sym}}^{n \times n}$, should generalize to $\mathcal{M}_{\mathcal{D}} \subseteq \mathbb{R}_{\text{sym}}^{n \times n}$):

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[K. (ta)] algebraic argument based on graph recursions and removable singularities

Spectral Functions — Inherited Differentiability

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Subsequent arguments are around matrices with distinct eigenvalues, where differentiability is given by analyticity of eig° .

Definition

$$\nabla^d F(X)[\Delta^{(1)}, \dots, \Delta^{(d)}] := \frac{\partial^d}{\partial t_1 \dots \partial t_d} F(X + t_1 \Delta^{(1)} + \dots + t_d \Delta^{(d)}) \Big|_{t_1, \dots, t_d=0}$$

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For invariances $\mathcal{I} \in \text{Lin}(\mathbb{K}^{n \times n}, \mathbb{K}^{n \times n})$ of F :

$$\nabla^d F(X)[\Delta^{(1)}, \dots, \Delta^{(d)}] = \nabla^d F(\mathcal{I}(X))[\mathcal{I}(\Delta^{(1)}), \dots, \mathcal{I}(\Delta^{(d)})]$$

In particular:

$$\nabla^d F(T \Lambda T^{-1})[\Delta^{(1)}, \dots, \Delta^{(d)}] = \nabla^d F(\Lambda)[T^{-1} \Delta^{(1)} T, \dots, T^{-1} \Delta^{(d)} T]$$

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Lemma ([K. (ta)])

$$\nabla^d F(\Lambda)[e_{i_1 j_1}, \dots, e_{i_d j_d}] = 0$$

if there is $\ell \in [n]$ for which

$$\deg_\ell^{(\text{out})} := |\{\mu \mid i_\mu = \ell\}| \neq |\{\mu \mid j_\mu = \ell\}| =: \deg_\ell^{(\text{in})}.$$

Proof: test $\mathcal{I}(X) := H_\ell X H_\ell^{-1}$, for $H_\ell = I + (\xi - 1)e_{\ell\ell}$ for certain root of unity ξ

Invariance based Euler Graph Relation — Derivative Map $\Theta_f(\mathcal{G})$

Definition

For a directed pseudograph (multigraph with loops allowed),

$$\mathcal{G} = \{i_1 \rightarrow j_1, \dots, i_d \rightarrow j_d\}_\#, \quad i, j \in [n]^d$$

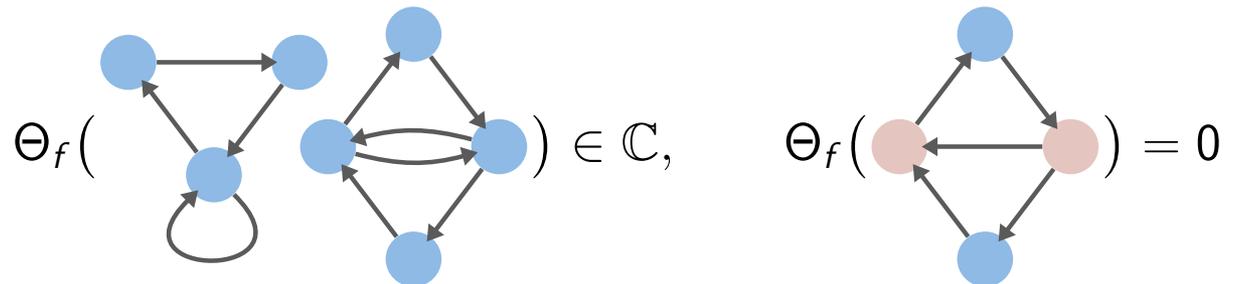
let

$$\Theta_f(\mathcal{G}) := \nabla^d F(\Lambda)[e_{i_1 j_1}, \dots, e_{i_d j_d}]$$

Corollary ([K. (ta)])

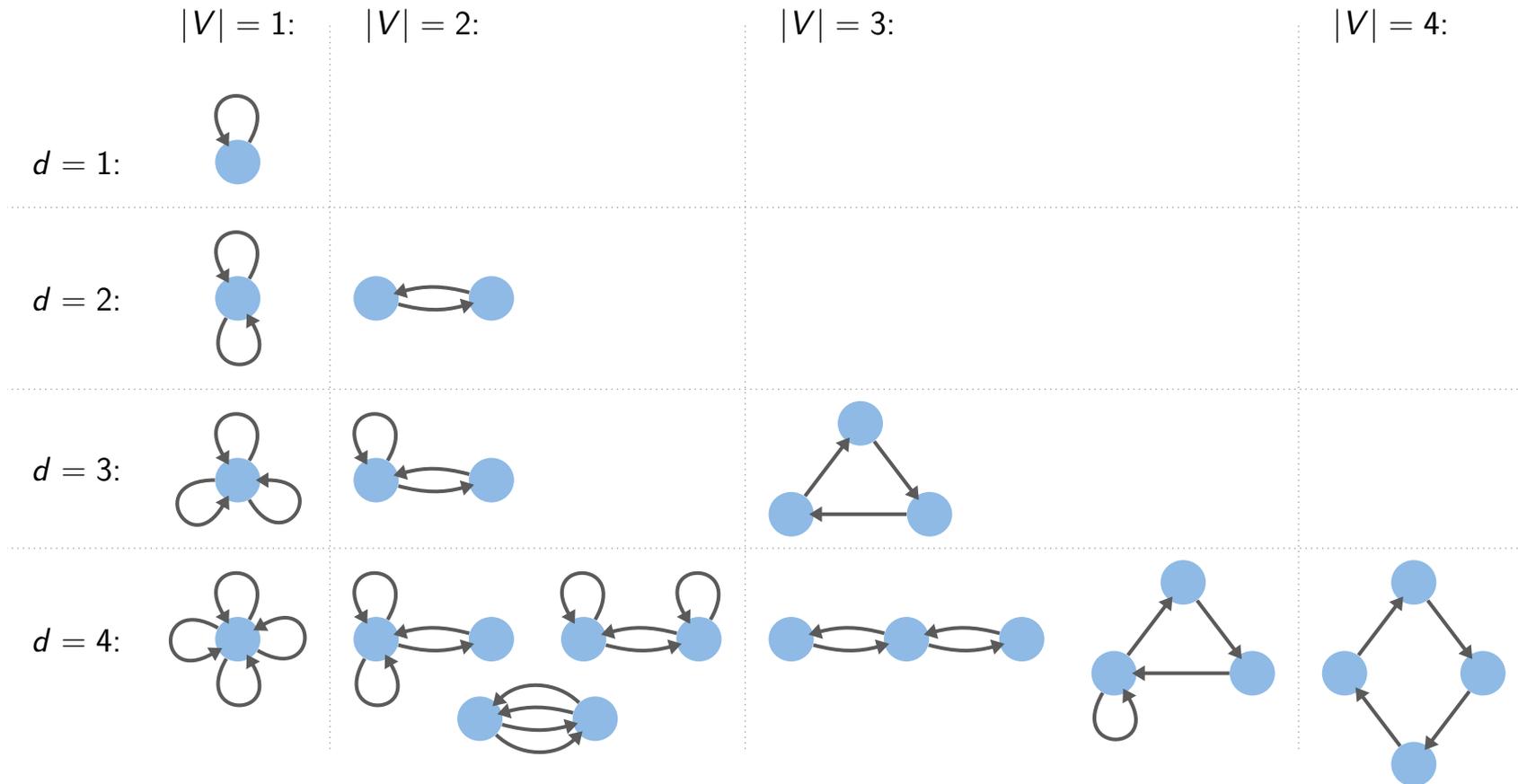
If \mathcal{G} is not a union of Euler graphs, then $\Theta_f(\mathcal{G}) = 0$.

union of Euler graphs:
(in-deg = out-deg)



Invariance based Euler Graph Relation — Classes of Euler Graphs

All equivalence classes of connected Euler (pseudo-)graphs up to 4 edges:



Invariance based Euler Graph Relation — Differentiation of Conjugation

For map of invariances $\mathcal{I} : (-\delta, \delta) \rightarrow \text{Lin}(\mathbb{K}^{n \times n}, \mathbb{K}^{n \times n})$ with $\mathcal{I}_0 = \text{id}$:

$$\begin{aligned} & - \nabla^d F(X)[\Delta^{(1)}, \dots, \Delta^{(d-1)}, \mathcal{I}'_0(X)] \\ & = \nabla^{d-1} F(X)[\mathcal{I}'_0(\Delta^{(1)}), \Delta^{(2)}, \dots] + \dots + \nabla^{d-1} F(X)[\dots, \Delta^{(d-2)}, \mathcal{I}'_0(\Delta^{(d-1)})]. \end{aligned}$$

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Use a basic idea of Lie group theory (turns out: generalizing ^[M.Šilhavý '00]),

$$\mathcal{I}_\alpha(X) := T_\alpha X T_\alpha^{-1}, \quad T_0 = I \quad \Rightarrow \quad \mathcal{I}'_0(E) = [T'_0, E] = T'_0 E - E T'_0$$

Inside an identity matrix, set

$$T_\alpha|_{(i,j) \times (i,j)} = \begin{pmatrix} \cos(\alpha) & \sin(\alpha) \\ -s \sin(\alpha) & \cos(\alpha) \end{pmatrix}, \quad s \in \{0, 1\}, \quad T'_0 = e_{ij} - s e_{ji}.$$

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For k and $i \neq j$:

$$\begin{aligned} [e_{ij} - se_{ji}, \Lambda] &= (\lambda_i - \lambda_j)(e_{ij} + se_{ji}) \\ [e_{ij} - se_{ji}, e_{jk} + se_{kj}] &= e_{ik} + se_{ki}, \quad [e_{ij} - se_{ji}, e_{ji} + se_{ij}] = (1 + s)(e_{ii} - e_{jj}) \end{aligned}$$

Invariance based Euler Graph Relation — Recursion

Theorem ([K. (ta)])

Let $r = i \rightarrow j \in \mathcal{G}$, $i \neq j$. Then

$$\begin{aligned}
 & (\lambda_i - \lambda_j) \cdot \Theta_f(\mathcal{G}) \\
 = & \sum_{e \in \mathcal{G} : e=j \rightarrow k} \Theta_f(\mathcal{G} \setminus \{r, e\} \cup \{i \rightarrow k\}) - \sum_{e \in \mathcal{G} : e=k \rightarrow i} \Theta_f(\mathcal{G} \setminus \{r, e\} \cup \{k \rightarrow j\})
 \end{aligned}$$

Example

One choice of r :

$$(\lambda_i - \lambda_j) \cdot \Theta_f(\text{graph}) = \Theta_f(\text{graph}_1) + \Theta_f(\text{graph}_2) - \Theta_f(\text{graph}_3)$$

Example

With another choice:

$$\begin{aligned}
 & (\lambda_i - \lambda_j) \cdot \Theta_f \left(\begin{array}{c} \bullet \\ \nearrow \quad \searrow \\ j \quad i \\ \nwarrow \quad \swarrow \\ \bullet \end{array} \right) \\
 = & \Theta_f \left(\begin{array}{c} \bullet \quad k \\ \nearrow \quad \searrow \\ \bullet \quad i \\ \nwarrow \quad \swarrow \\ \bullet \end{array} \right) + \Theta_f \left(\begin{array}{c} \bullet \\ \nearrow \quad \searrow \\ \bullet \quad i, k \\ \nwarrow \quad \swarrow \\ \bullet \end{array} \right) - \Theta_f \left(\begin{array}{c} \bullet \\ \nearrow \quad \searrow \\ j, k \quad \bullet \\ \nwarrow \quad \swarrow \\ \bullet \end{array} \right) - \Theta_f \left(\begin{array}{c} \bullet \quad k \\ \nearrow \quad \searrow \\ j \quad \bullet \\ \nwarrow \quad \swarrow \\ \bullet \end{array} \right)
 \end{aligned}$$

derivatives of F (edges and loops) reduce to those of f (loops only)
(as in chain rule evaluation for $f^\circ \circ \text{eig}^\circ$)

Euler Graph Relation — Differential Polynomial $\theta_1(\mathcal{G})$

Definition

Capture derivatives in

$$\theta(\mathcal{G}) \in \mathbb{C}\left[\left\{\frac{1}{\lambda_i - \lambda_j}\right\}_{i \neq j}\right][\delta_1, \dots, \delta_n] \quad \text{s.t.} \quad \theta(\mathcal{G})\left(\frac{\partial}{\partial \lambda_1}, \dots, \frac{\partial}{\partial \lambda_n}\right)f = \Theta_f(\mathcal{G})$$

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Example

The differential polynomial of the loop-only graph

$$\theta\left(\begin{array}{c} \text{loop on node 1} \\ \text{loop on node 2} \end{array}\right) = \delta_1^2 \delta_2$$

corresponds to

$$\Theta_f\left(\begin{array}{c} \text{loop on node 1} \\ \text{loop on node 2} \end{array}\right) = \nabla^3 F(\Lambda)[\mathbf{e}_{11}, \mathbf{e}_{11}, \mathbf{e}_{22}] = \frac{\partial^2}{\partial X_{11}^2} \frac{\partial}{\partial X_{22}} F(\Lambda) \stackrel{f=F \circ \text{diag}}{=} \frac{\partial^2}{\partial \lambda_1^2} \frac{\partial}{\partial \lambda_2} f(\lambda)$$

Euler Graph Relation — The Simple Cycle

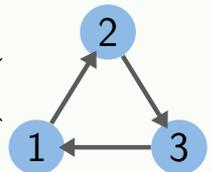
Lemma

For a canonical, simple cycle:

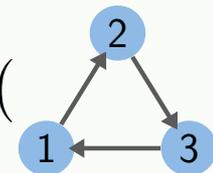
$$\mathcal{G} = \{1 \rightarrow 2, \dots, d \rightarrow 1\}, \quad \theta(\mathcal{G}) = \sum_{i=1}^d \left(\prod_{j=1, j \neq i}^d \frac{1}{\lambda_i - \lambda_j} \right) \cdot \delta_i.$$

For trace functions $f_g(x) = g(x_1) + \dots + g(x_n)$ follows $\Theta_{f_g}(\mathcal{G}) = [\lambda_1, \dots, \lambda_d]g'$.

Example


$$\theta(\text{cycle}) = \frac{1}{\lambda_2 - \lambda_1} \frac{1}{\lambda_3 - \lambda_1} \delta_1 + \frac{1}{\lambda_1 - \lambda_2} \frac{1}{\lambda_3 - \lambda_2} \delta_2 + \frac{1}{\lambda_1 - \lambda_3} \frac{1}{\lambda_2 - \lambda_3} \delta_3$$

and for trace functions,


$$\Theta_{f_g}(\text{cycle}) = [\lambda_1, \lambda_2, \lambda_3]g'$$

Euler Graph Relation — Linear Part $\theta_1(\mathcal{G})$ of Differential Polynomials

Lemma

The linear part $\theta_1(\mathcal{G}) := \theta(\mathcal{G})|_{\text{degree } 1 \text{ terms}}$ is locally given via

$$\theta_1(\mathcal{G}) = \Theta_{f_{g^{(1)}}}(\mathcal{G}) \cdot \delta_1 + \dots + \Theta_{f_{g^{(n)}}}(\mathcal{G}) \cdot \delta_n$$

where

$$f_{g^{(i)}}(\lambda) = \lambda_i, \quad g^{(i)}(x) = \chi_{U_{\lambda_i}}(x) \cdot x, \quad F_{g^{(i)}}(X) = \text{eig}_i^\circ(X)$$

where $\chi_{U_{\lambda_i}}$ are the indicator functions on sufficiently small neighborhoods of λ_i

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The graph recursion holds true for

- ▶ derivatives $\Theta_f(\mathcal{G})$,
- ▶ differential polynomial $\theta(\mathcal{G})$,
- ▶ its linear part $\theta_1(\mathcal{G})$,

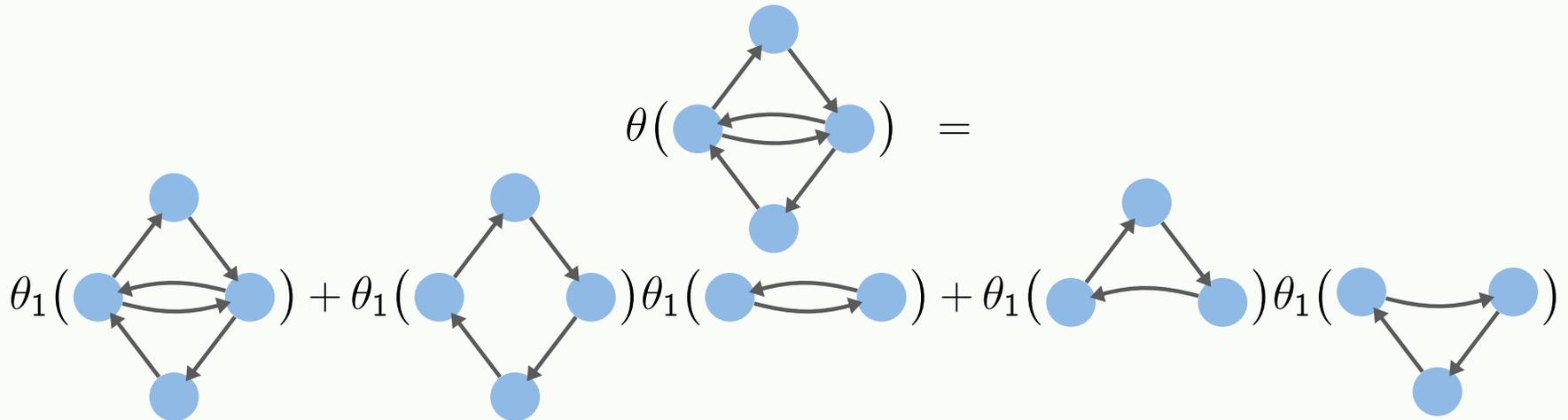
for distinct eigenvalues λ (though generalizes in certain ways to all)

Euler Graph Relation — Factorization of Differential Polynomial

Theorem ([K. (ta)])

Factorization via,
$$\theta(\mathcal{G}) = \sum_{\cup_{\nu} \text{Euler}_{\nu} = \mathcal{G}} \prod_{\nu} \theta_1(\text{Euler}_{\nu})$$

Example

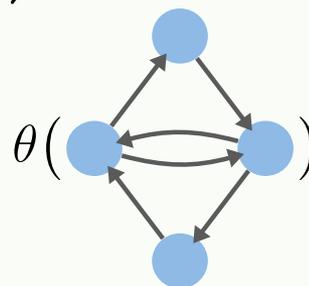


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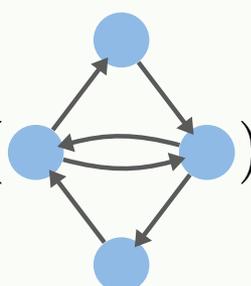
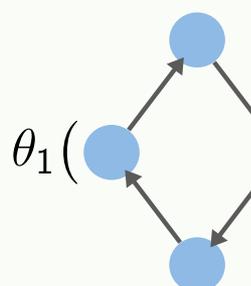
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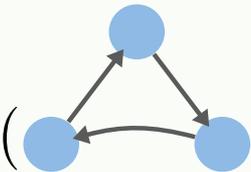
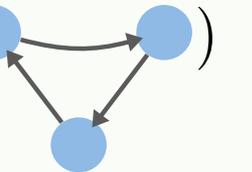
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Example

$$\nabla^6 F(\Lambda)[e_{ij} : i \rightarrow j \in \text{graph}]$$


$$\theta(\text{graph}) =$$

$$\underbrace{\theta_1(\text{graph}_1)}_{\nabla f(\lambda)[\text{coeffs } \theta_1(\text{graph}_1)]} + \underbrace{\theta_1(\text{graph}_2)}_{\nabla^2 f(\lambda)[\text{coeffs } \theta_1(\text{graph}_2), \text{coeffs } \theta_1(\text{graph}_3)]} \theta_1(\text{graph}_3) + \underbrace{\theta_1(\text{graph}_4)}_{\nabla^2 f(\lambda)[\text{coeffs } \theta_1(\text{graph}_4), \text{coeffs } \theta_1(\text{graph}_5)]} \theta_1(\text{graph}_5)$$



Inflation to Euler Circuits — Vertex Exchange

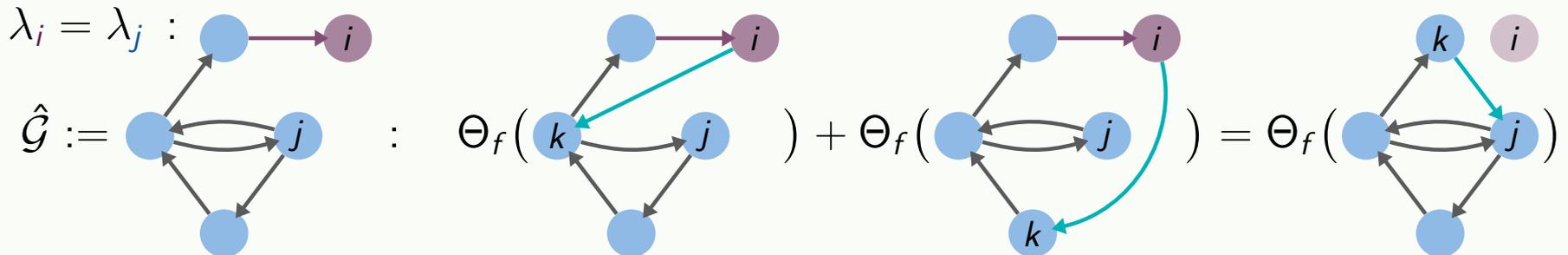
Corollary ([K. (ta)])

Let $i \neq j$. If $\lambda_i = \lambda_j$, then

$$\sum_{e \in \mathcal{G} : e=j \rightarrow k} \Theta_f(\mathcal{G} \setminus \{j \rightarrow k\} \cup \{i \rightarrow k\}) = \sum_{e \in \mathcal{G} : e=k \rightarrow i} \Theta_f(\mathcal{G} \setminus \{k \rightarrow i\} \cup \{k \rightarrow j\}).$$

(At least) for trace functions, we can augment $i := n + 1 \notin \mathcal{G}$:

Example



Inflation to Euler Circuits — Vertex Exchange

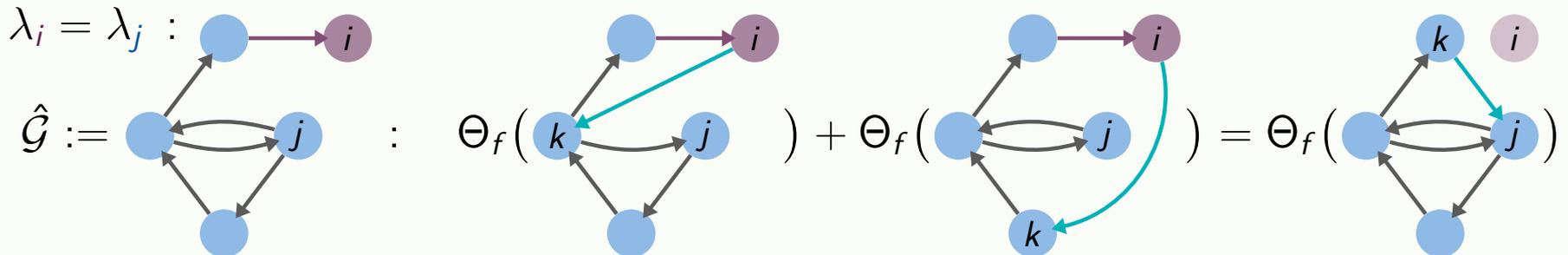
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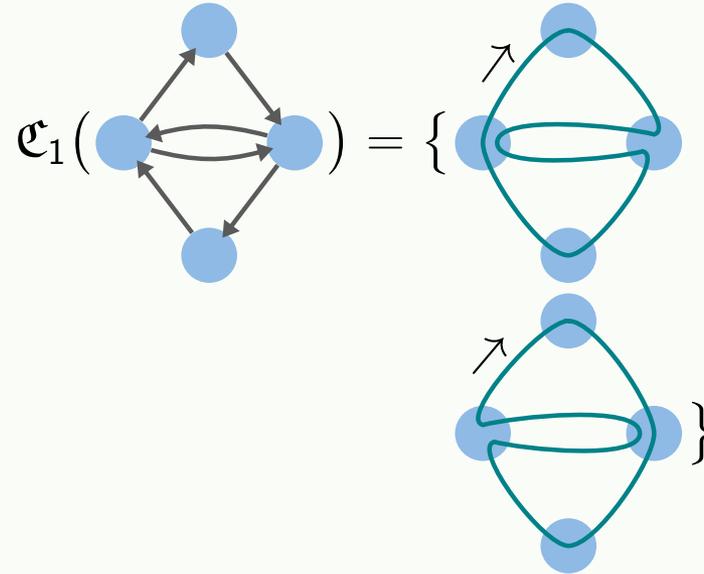
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Iterating this encounters Euler circuits! $=: \mathfrak{C}_1(\mathcal{G})$

Example



Theorem ([K. (ta)])

Let

$$\mathcal{G} = \{i_1 \rightarrow j_1, \dots, i_d \rightarrow j_d\} \quad \text{s.t.} \quad \Theta_f(\mathcal{G}) = \nabla^d F(\Lambda)[e_{i_1 j_1}, \dots, e_{i_d j_d}]$$

For trace functions,

$$\Theta_{f_g}(\mathcal{G}) = |\mathfrak{C}_1(\mathcal{G})| \cdot [\lambda_{i_1}, \dots, \lambda_{i_d}] g'$$

For the linear part of the differential polynomial,

$$\theta_1(\mathcal{G}) = |\mathfrak{C}_1(\mathcal{G})| \cdot \sum_{k=1}^n [\lambda_{i_1}, \dots, \lambda_{i_d}] \chi_{U_{\lambda_k}} \cdot \delta_k$$

Proof: inflate graph to sums of simple cycles / use $\theta_1(\mathcal{G})$ relation

- ▶ $\Theta_{f_g}(\mathcal{G})$ identity valid for all λ (under Hermite scheme)
- ▶ $\theta_1(\mathcal{G})$ identity **not** compatible with coinciding eigenvalues (but uses $\chi'_{U_{\lambda_k}}(\lambda_k) = 0$)

Theorem

For trace functions F_g (as for instance also by [F.Hansen '06, N.Highem '08])

$$\nabla^d F_g(\Lambda)[\Delta^{(1)}, \dots, \Delta^{(d)}] = \frac{1}{d} \sum_{m \in [n]^d} [\lambda_m] g' \cdot \sum_{\pi \in S_d} \Delta_{m_1 m_2}^{(\pi(1))} \cdot \Delta_{m_2 m_3}^{(\pi(2))} \cdots \Delta_{m_d m_1}^{(\pi(d))}$$

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as well as (generalizing [H.S.Sendov '07] to non-symmetric matrices)

$$\nabla^d F_g(\Lambda)[\Delta^{(1)}, \dots, \Delta^{(d)}] = \sum_{m \in [n]^d} [\lambda_m] g' \cdot \sum_{\tau \in S_d : \tau \text{ cyclic}} \prod_{\mu \in [d]} \Delta_{m_\mu, m_{\tau(\mu)}}^{(\mu)}$$

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Corollary

For an isolated eigenvalue $E_1^{(0)} := \lambda_1$,

$$F(X) := \text{eig}_1^\circ(\Lambda + t\Delta) = E_1^{(0)} + \sum_{d \in \mathbb{N}} t^d E_1^{(d)}$$

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Then $E_1^{(1)} = \Delta_{11}$ and (as also by [T.Kato '95])

$$E_1^{(d)} = \frac{1}{d!} \nabla^d \text{eig}_1^\circ(\Lambda)[\Delta, \dots, \Delta] = \frac{1}{d} \sum_{m \in [n]^d} \Delta_{m_1 m_2} \dots \Delta_{m_d m_1} \cdot [\lambda_m] \chi_{U_{\lambda_1}}$$

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with $S^0 := e_{11}$ and $S^q := \text{diag}(0, \frac{1}{\lambda_1 - \lambda_2}, \dots, \frac{1}{\lambda_1 - \lambda_n})^q$, $q \in \mathbb{N}$.

Proof: combinatorics and symmetric polynomial expression for $[\lambda_m] \chi_{U_{\lambda_1}}$
 [T.Kato '95] contour-integral around expanded trace-resolvent

Reverse Applications — Counting Euler Circuits in Undirected Graphs

Use

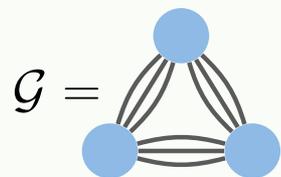
$$\Theta_{f_{\log}}(\mathcal{G}_{\text{undir}})(\mathbf{1}) = \pm |\mathfrak{e}_1(\mathcal{G}_{\text{undir}})|, \quad \text{trace log} = \log \det, \quad \mathcal{X} := \begin{pmatrix} 1 & X_{12} & \dots & X_{1n} \\ X_{12} & 1 & \dots & X_{2n} \\ \vdots & \vdots & & \vdots \\ X_{1n} & X_{2n} & \dots & 1 \end{pmatrix}$$

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Example



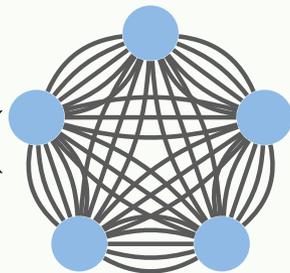
$$\begin{aligned} \mathcal{G} & : \log \det(\mathcal{X}) \xrightarrow{\left(\frac{\partial^3}{\partial X_{12}^3} u\right) |_{X_{12}=0}} \frac{-4X_{13}X_{23}(4X_{13}^2X_{23}^2 - 3X_{13}^2 - 3X_{23}^2 + 3)}{(X_{13}^2 + X_{23}^2 - 1)^3} \\ & \xrightarrow{\left(\frac{\partial^3}{\partial X_{23}^3} u\right) |_{X_{23}=0}} \frac{-48X_{23}(2X_{23}^2 + 3)}{(X_{23} - 1)^3(X_{23} + 1)^3} \xrightarrow{\left(\frac{\partial^3}{\partial X_{13}^3} u\right) |_{X_{13}=0}} 3168 = |\mathfrak{C}_1(\mathcal{G})|. \end{aligned}$$

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Example


$$\mathfrak{e}_1(\text{K}_5) = 15414462840434593737631334400000$$

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$$\mathfrak{e}_1 \left(\begin{array}{c} \bullet \\ \bullet \quad \bullet \\ \bullet \quad \bullet \quad \bullet \\ \bullet \quad \bullet \quad \bullet \quad \bullet \\ \bullet \quad \bullet \quad \bullet \quad \bullet \quad \bullet \end{array} \right) = 30480384$$

Singular Value Functions —

Definition

An *sv function* maps a matrix to a symmetric **and even** function of its singular values,

$$F = f \circ \text{sv} : M \subseteq \mathbb{R}^{n \times n} \rightarrow \mathbb{R}, \quad f : \text{sv}(M) \subseteq \mathbb{R}^n \rightarrow \mathbb{R} \text{ (sym \& even)}$$

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- ▶ here, not every Euler circuit yields same summand
- ▶ for graph-independent expressions for even trace functions, may also use

$$\nabla^d F_g^{(\text{sv})}(\Sigma)[\Delta^{(1)}, \dots, \Delta^{(d)}] = \frac{1}{2} \nabla^d F_g^{(\text{eig})}(\text{diag}(-\Sigma, \Sigma))[\Omega^{(1)}, \dots, \Omega^{(d)}]$$

$$\Omega^{(\mu)} := \frac{1}{2} \begin{pmatrix} -\Delta^{(\mu)} - \Delta^{(\mu)T} & -\Delta^{(\mu)} + \Delta^{(\mu)T} \\ \Delta^{(\mu)} - \Delta^{(\mu)T} & \Delta^{(\mu)} + \Delta^{(\mu)T} \end{pmatrix}$$

Open ends of this approach:

- ▶ simplify derivatives at low-rank and non-symmetric matrices
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- ▶ generalize formulas to non-trace functions for non-distinct eigenvalues

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Thank you for your attention!

Slide overview:

Spectral Functions

- Introduction
- About Symmetry
- Inherited Analyctiy
- Inherited Differentiability

Invariance based Euler Graph Relation

- Diagonalization and Degree Condition
- Derivative Map $\Theta_f(\mathcal{G})$
- Classes of Euler Graphs
- Differentiation of Conjugation
- Recursion

Euler Graph Relation

- Differential Polynomial $\theta_1(\mathcal{G})$

The Simple Cycle

- Linear Part $\theta_1(\mathcal{G})$ of Differential Polynomials
- Factorization of Differential Polynomial

Inflation to Euler Circuits

- Vertex Exchange
- Euler Circuits
- Graph-Independent Expressions

Graph-Independent Expressions

- Trace Functions
- Perturbation Theory

Reverse Applications

- Counting Euler Circuits in Undirected Graphs

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