

Hodge decomposition in data analysis

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this talk

- prepared for a different audience
- according to AMS short course manual:
 - ▶ the idly curious, knowing little or nothing specific of the field beyond a layman's or graduate student's familiarity
 - ▶ peripheralists, who have read a few articles, perhaps dabbled once or twice in the field, and would like to have a perspective of the field presented on a silver platter
 - ▶ young specialists and prospective teachers, who want to make sure they see the forest for the trees, and haven't missed something significant
- objectives: present as simply as possible
 - ▶ two ideas
 - ★ cohomology
 - ★ Hodge decomposition
 - ▶ two applications
 - ★ ranking: web search, recommendation systems, crowd sourcing
 - ★ game theory: ad auction, happiness index, social networks

Cohomology for Pedestrians

Laplace equation

- Laplace or homogeneous Poisson equation in \mathbb{R}^3 :

$$\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} + \frac{\partial^2 f}{\partial z^2} = 0$$

- more generally

$$\Delta f = \nabla^2 f = \nabla \cdot \nabla f = \operatorname{div} \operatorname{grad} f = \sum_{i=1}^n \frac{\partial^2 f}{\partial x_i^2}$$

electrostatics: electric potential in free space with no charge

fluid mechanics: velocity potential of incompressible fluid

thermal conduction: stationary heat equation without a heat source

Laplace equation: 0-cohomology

- Laplace or homogeneous Poisson equation:

$$\begin{cases} \Delta f = 0 & \text{in } \Omega \\ f = g & \text{on } \partial\Omega \end{cases} \quad \text{or} \quad \begin{cases} \Delta f = 0 & \text{in } \Omega \\ \partial f / \partial n = g & \text{on } \partial\Omega \end{cases}$$

- tells us about the topology and geometry of Ω
- solution $f : \Omega \rightarrow \mathbb{R}$ called **harmonic function**
- 0-cohomology is the study of solutions to Laplace equation with no boundary conditions
- 0-cohomology class is harmonic function
- 0-cohomology group is set of all harmonic functions

vector Laplace equation: 1-cohomology

- vector Laplace or homogeneous vector Poisson equation in \mathbb{R}^3 :

$$\begin{cases} -\operatorname{grad} \operatorname{div} f + \operatorname{curl} \operatorname{curl} f = 0 & \text{in } \Omega \\ f \cdot n = 0, \quad \operatorname{curl} f \times n = 0 & \text{on } \partial\Omega \end{cases}$$

- Helmholtz operator** or **vector Laplacian** in \mathbb{R}^3

$$\Delta_1 f = \operatorname{curl} \operatorname{curl} f - \operatorname{grad} \operatorname{div} f = \nabla(\nabla \cdot f) - \nabla \times (\nabla \times f) = \nabla^2 f$$

- solution $f : \Omega \rightarrow \mathbb{R}^3$ is vector field on Ω , call this **harmonic 1-form**
- 1-cohomology is the study of solutions to vector Laplace equation with no boundary conditions
- 1-cohomology class is harmonic 1-form
- 1-cohomology group is set of all harmonic 1-form

cohomology for pedestrians

- 0th cohomology classes are solutions to scalar Laplace equation

$$H^0(\Omega) = \ker(\Delta) = \{f : \Delta f = 0\}$$

- 1th cohomology classes are solutions to vector Laplace equation

$$H^1(\Omega) = \ker(\Delta_1) = \{f : \Delta_1 f = 0\}$$

- for $k = 0$

$$\Delta_0 = \text{div grad}$$

is our usual Laplace operator or scalar Laplacian Δ

- for $k = 1$

$$\Delta_1 = -\text{grad div} + \text{curl curl}$$

is our usual Helmholtz operator or vector Laplacian

- k th cohomology classes: use 'higher-order Laplacians'

$$\Delta_k = \delta_{k-1} \delta_{k-1}^* + \delta_k^* \delta_k$$

three approaches

ordinary: $\delta_k : C^k(X) \rightarrow C^{k+1}(X)$ coboundary operators,

$$H^k(X) = \ker(\delta_k) / \text{im}(\delta_{k-1})$$

generalized: $\{(E_k, \varepsilon_k) \mid k \in \mathbb{Z}\}$ spectrum,

$$H^k(X) = [X, E_k]$$

harmonic: $\Delta_k = \delta_{k-1}\delta_{k-1}^* + \delta_k^*\delta_k$ combinatorial Laplacian,

$$H^k(X) = \ker(\Delta_k)$$

why haven't you seen the last one before?

harmonic approach

- cons
- not functorial
 - metric dependent
 - doesn't work over rings
 - doesn't work over fields of positive characteristics
- pros
- each cohomology class has unique harmonic representative
 - works in noisy setting: eigenfuctions of Δ_k with small eigenvalues [De Silva, 2006]
 - accessible to practitioners
 - comes with a Hodge decomposition

Hodge decomposition

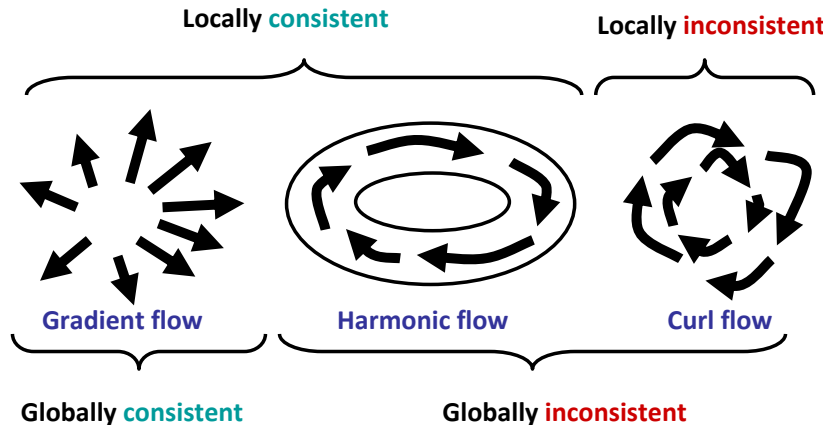


Figure : cartoon courtesy of Pablo Parrilo

easy to apply

- fluid mechanics

$$\text{fluid flow} = \text{irrotational} \oplus \text{solenoidal} \oplus \text{harmonic}$$

- ranking

$$\begin{aligned} \text{pairwise ranking} = \\ \text{consistent} \oplus \text{locally inconsistent} \oplus \text{globally inconsistent} \end{aligned}$$

- games

$$\begin{aligned} \text{multiplayer game} = \\ \text{potential game} \oplus \text{nonstrategic game} \oplus \text{harmonic game} \end{aligned}$$

Russell Crowe's problem

$$V = \{F : \mathbb{R}^3 \setminus X \rightarrow \mathbb{R}^3 \mid \nabla \times F = 0\}; \quad W = \{F = \nabla g\}; \quad \dim(V/W) = ?$$

Netflix Problem

ranking problems

- **static ranking (Google problem)**
 - ▶ alternatives: football teams, websites
 - ▶ **one voter**: entire season of games, hyperlink structure of WWW
 - ▶ **one ranking**: number of matches won by each team, PageRank of each website
 - ▶ no paradox, impossibility, chaos, NP-hardness
- **collaborative filtering (the better known Netflix problem)**
 - ▶ alternatives: movies, drugs
 - ▶ **many voters**: movie viewers, patients
 - ▶ **many rankings**: ideally one for each viewer, patient
 - ▶ no paradox, impossibility, chaos, NP-hardness
- **rank aggregation (our Netflix problem)**
 - ▶ alternatives: colleges, candidates
 - ▶ **many voters**: academics surveyed, electorate
 - ▶ **one ranking**: order all alternatives globally
 - ▶ Condorcet's paradox, Arrow's impossibility, McKelvey's chaos, NP-hard

the Netflix problem in this talk

Watch Instantly | Browse DVDs | Your Queue | ★ Suggestions For You

Genres ▾ | New Releases | **Netflix Top 100** | Critics' Picks | Award Winners

Search: Movies, TV shows, actors, directors, genres

Netflix Top 100

You have 217 Suggestions from 105 ratings.

Rank	Action	Title	Rating
1.	Add	The Blind Side	4.5
2.	Add	Crash	4.0
3.	Add	The Bucket List	4.0
4.	Add	The Curious Case of Benjamin Button	4.0
5.	Add	The Hurt Locker	4.0
6.	Add	The Departed	4.0
7.	Add	Sherlock Holmes	4.0
8.	Add	Inception	4.0
9.	Add	Iron Man	4.0
10.	Add	No Country for Old Men	4.0
11.	Add	Date Night	3.5
12.	Add	Up in the Air	3.5

Helpful Tip
◀ **Seen any of these movies?**
★★★★☆
Rate movies you've seen before so we can recommend movies you haven't!

Give FREE rentals!
Tell a friend

ⓘ Add this page to your favorite web portal or RSS reader.
[Learn more about RSS...](#)

rank aggregation

- many voters, each rated a few alternatives, want global ranking
- averaging over scores doesn't work: one movie receives one 5☆ and no other ratings, another receives 10,000 5☆ and one 4☆
- should be invariant under monotone transformation:

$$1\star, \dots, 5\star \longrightarrow 0\star, \dots, 4\star$$

- **basic unit of ranking:** pairwise comparison or pairwise ranking
- take average over pairwise rankings instead, get $Y \in \mathbb{R}^{17770 \times 17770}$
- for Netflix data, user-product rating matrix $Z \in \mathbb{R}^{480189 \times 17770}$ has 98.82% missing values, Y has 0.22% missing values

linear model: average score difference between i and j over all who have rated both,

$$y_{ij} = \frac{\sum_h (z_{hj} - z_{hi})}{\#\{h \mid z_{hi}, z_{hj} \text{ exist}\}}$$

invariant under translation

averaging over pairwise rankings

log-linear model: logarithmic average score ratio of positive scores,

$$y_{ij} = \frac{\sum_h (\log z_{hj} - \log z_{hi})}{\#\{h \mid z_{hi}, z_{hj} \text{ exist}\}}$$

invariant up to a multiplicative constant

linear probability model: probability j preferred to i in excess of purely random choice,

$$y_{ij} = \Pr\{h \mid z_{hj} > z_{hi}\} - \frac{1}{2}$$

invariant under monotone transformation

Bradley-Terry model: logarithmic odd ratio (logit),

$$y_{ij} = \log \frac{\Pr\{h \mid z_{hj} > z_{hi}\}}{\Pr\{h \mid z_{hj} < z_{hi}\}}$$

invariant under monotone transformation

difficulties with rank aggregation

- **Condorcet's paradox:** majority vote intransitive $i \succ j \succ k \succ i$ [Condorcet, 1785]
- **Arrow/Sen's impossibility:** any sufficiently sophisticated preference aggregation must exhibit intransitivity [Arrow, 1950], [Sen, 1970]
- **McKelvey/Saari's chaos:** almost every possible ordering can be realized by a clever choice of the order in which decisions are taken [McKelvey, 1979], [Saari, 1989]
- **Kemeny optimal is NP-hard:** even with just 4 voters [Dwork–Kumar–Naor–Sivakumar, 2001], quadratic assignment problem [Cook–Kress, 1984]
- **empirical evidence:** lack of consensus common in group decision making (e.g. US congress)

what we want

ordinal: intransitivity, $i \succ j \succ k \succ i$

cardinal: inconsistency, $X_{ij} + X_{jk} + X_{ki} \neq 0$

- want global ranking of alternatives if a reasonable one exists
- want certificate of reliability to quantify validity of global ranking
- if no meaningful global ranking, analyze nature of inconsistencies

Graph Theoretic Hodge Theory

graphs

- $G = (V, E)$ undirected graph
- V vertices
- $E \subseteq \binom{V}{2}$ edges
- $T \subseteq \binom{V}{3}$ triangles or **3-cliques**, i.e.,

$$\{i, j, k\} \in T \quad \text{iff} \quad \{i, j\}, \{j, k\}, \{k, i\} \in E$$

- more generally $K_k \subseteq \binom{V}{k}$ **k -cliques**, i.e.,

$$\{i_1, \dots, i_k\} \in K_k \quad \text{iff} \quad \text{it is a complete subgraph of } G$$

- nonempty family K of finite subsets of a set G is **abstract simplicial complex** if for every set X in K , every $Y \subseteq X$ also belongs to K
- $K(G)$ **clique complex** of a graph G is an abstract simplicial complex

functions on graphs

vertex functions: $s : V \rightarrow \mathbb{R}$

edge flows: $X : E \rightarrow \mathbb{R}$,

$$X(i, j) = -X(j, i) \quad \text{for all } i, j$$

triangular flows: $\Phi : T \rightarrow \mathbb{R}$,

$$\begin{aligned} \Phi(i, j, k) &= \Phi(j, k, i) = \Phi(k, i, j) \\ &= -\Phi(j, i, k) = -\Phi(i, k, j) = -\Phi(k, j, i) \quad \text{for all } i, j, k \end{aligned}$$

physics: s, X, Φ potential, alternating vector/tensor field

topology: s, X, Φ 0-, 1-, 2-cochain

ranking: s scores/utility, X pairwise rankings, Φ triplewise rankings

operators on functions on graphs

gradient: $\text{grad} : L^2(V) \rightarrow L^2(E)$,

$$(\text{grad } s)(i, j) = s_j - s_i$$

curl: $\text{curl} : L^2(E) \rightarrow L^2(T)$,

$$(\text{curl } X)(i, j, k) = X_{ij} + X_{jk} + X_{ki}$$

divergence: $\text{div} : L^2(E) \rightarrow L^2(V)$,

$$(\text{div } X)(i) = \sum_{j: \{i, j\} \in E} w_{ij} X_{ij}$$

graph Laplacian: $\Delta_0 : L^2(V) \rightarrow L^2(V)$,

$$\Delta_0 = \text{div grad}$$

graph Helmholtzian: $\Delta_1 : L^2(E) \rightarrow L^2(E)$,

$$\Delta_1 = -\text{grad div} + \text{curl}^* \text{curl}$$

generalization: cochains

- K abstract simplicial complex with vertex set V
- alternating functions on $k + 1$ arguments, i.e., k -forms or k -cochains:

$$C^k(K; \mathbb{R}) = \{u : K_{k+1} \rightarrow \mathbb{R} \mid u(i_{\sigma(0)}, \dots, i_{\sigma(k)}) = \text{sign}(\sigma)u(i_0, \dots, i_k)\}$$

for $(i_0, \dots, i_k) \in K_{k+1}$ and $\sigma \in \mathfrak{S}_{k+1}$

- most interesting for us $K = K(G)$, clique complex of graph G
- may put metrics/inner products on $C^k(K(G); \mathbb{R})$
- e.g. following metric on 1-forms, is useful for imbalanced ranking data:

$$\langle w_{ij}, \omega_{ij} \rangle_D = \sum_{(i,j) \in E} D_{ij} w_{ij} \omega_{ij}$$

where

D_{ij} = number of voters who rated both i and j

generalization: coboundary maps

- k -coboundary maps $\delta_k : C^k(K; \mathbb{R}) \rightarrow C^{k+1}(K; \mathbb{R})$ are

$$(\delta_k u)(i_0, \dots, i_{k+1}) = \sum_{j=0}^{k+1} (-1)^{j+1} u(i_0, \dots, i_{j-1}, i_{j+1}, \dots, i_{k+1})$$

- **fundamental theorem of topology:** $\delta_{k+1}\delta_k = 0$
- for $k = 0$,

$$\begin{array}{c} C^0 \xrightarrow{\delta_0} C^1 \xrightarrow{\delta_1} C^2 \\ \text{global} \xrightarrow{\delta_0} \text{pairwise} \xrightarrow{\delta_1} \text{triplewise} \\ \text{global} \xleftarrow{\delta_0^*} \text{pairwise} \xleftarrow{\delta_1^*} \text{triplewise} \end{array}$$

- we have $\delta_1\delta_0(\text{global rankings}) = 0$, i.e.,
 - ▶ global rankings are transitive/consistent
 - ▶ no need to consider rankings beyond triplewise

combinatorial Laplacian and Hodge theory

k -dimensional **combinatorial Laplacian**, $\Delta_k : C^k \rightarrow C^k$ by

$$\Delta_k = \delta_{k-1}\delta_{k-1}^* + \delta_k^*\delta_k, \quad k > 0$$

call u a **harmonic form** if $\Delta_k u = 0$

Theorem (Hodge)

- 1 $H^k(K; \mathbb{R}) = \ker(\delta_k) / \text{im}(\delta_{k-1}) \cong \ker(\Delta_k)$
- 2 $C^k(K; \mathbb{R}) = \text{im}(\delta_{k-1}) \oplus \ker(\Delta_k) \oplus \text{im}(\delta_k^*)$
- 3 $\ker(\Delta_k) = \ker(\delta_k) \cap \ker(\delta_{k-1}^*)$

follows from fundamental theorem of topology and Fredholm alternative:

$$\mathbb{R}^n = \ker(A) \oplus \text{im}(A^*), \quad \mathbb{R}^m = \ker(A^*) \oplus \text{im}(A)$$

for $A \in \mathbb{R}^{m \times n}$

special case: Helmholtz decomposition

Theorem (Helmholtz decomposition for graphs)

$G = (V, E)$ graph. The space of edge flows, $C^1(K(G), \mathbb{R})$, admits an orthogonal decomposition into three subspaces

$$C^1(K(G), \mathbb{R}) = \text{im}(\text{grad}) \oplus \ker(\Delta_1) \oplus \text{im}(\text{curl}^*)$$

where

$$\ker(\Delta_1) = \ker(\text{curl}) \cap \ker(\text{div}).$$

Application to Ranking

X. Jiang, L.-H. Lim, Y. Yao, and Y. Ye, "Statistical ranking and combinatorial Hodge theory," *Math. Program.*, **127** (2011), no. 1, pp. 203–244.

Helmholtz decomposition applied to rankings

pairwise comparison graph $G = (V, E)$; V : set of alternatives, E : pairs of alternatives compared

Theorem (Helmholtz decomposition for pairwise rankings)

The space of pairwise rankings, $C^1(K(G), \mathbb{R})$, admits an orthogonal decomposition into three components

$$C^1(K(G), \mathbb{R}) = \text{im}(\text{grad}) \oplus \ker(\Delta_1) \oplus \text{im}(\text{curl}^*)$$

where

$$\ker(\Delta_1) = \ker(\text{curl}) \cap \ker(\text{div}).$$

our approach: HodgeRank

- Hodge decomposition of ranking:

$$\text{aggregate pairwise ranking} = \\ \text{consistent} \oplus \text{locally inconsistent} \oplus \text{globally inconsistent}$$

- consistent component gives global ranking
- total size of inconsistent components gives certificate of reliability
- local and global inconsistent components tell us about nature of inconsistencies
- quantifies Condorcet paradox, Arrow's impossibility, McKelvey chaos, etc
- numerical, not combinatorial, so not NP-hard

properties

- $\text{im}(\text{grad})$: pairwise rankings that are gradient of score functions, i.e., consistent or **integrable**
- $\text{ker}(\text{div})$: $\text{div } X(i)$ measures the **inflow-outflow sum** at i ; $\text{div } X(i) = 0$ implies alternative i is preference-neutral in all pairwise comparisons
- $\text{ker}(\text{curl})$: pairwise rankings with zero flow-sum along any triangle
- $\text{ker}(\Delta_1) = \text{ker}(\text{curl}) \cap \text{ker}(\text{div})$: **globally inconsistent** or harmonic rankings; no inconsistencies due to small loops of length 3, i.e., $a \succ b \succ c \succ a$, but inconsistencies along larger loops of lengths > 3
- $\text{im}(\text{curl}^*)$: **locally inconsistent** rankings; non-zero curls along triangles
- div grad is **vertex Laplacian**
- $\text{curl}^* \text{curl}$ is **edge Laplacian**

analyzing inconsistencies

- locally inconsistent rankings should be acceptable
 - ▶ inconsistencies in items ranked close together but not in items ranked far apart
 - ▶ ordering of 4th, 5th, 6th ranked items cannot be trusted but ordering of 4th, 50th, 600th ranked items can
 - ▶ e.g. no consensus for hamburgers, hot dogs, pizzas, and no consensus for caviar, foie gras, truffle, but clear preference for latter group
- globally inconsistent rankings ought to be rare

Theorem (Kahle, 2007)

Erdős-Rényi $G(n, p)$, n alternatives, comparisons occur with probability p , clique complex χ_G almost always have zero 1-homology, unless

$$\frac{1}{n^2} \ll p \ll \frac{1}{n}.$$

relates to Kemeny optimum

- ranking data live on **pairwise comparison graph** $G = (V, E)$; V : set of alternatives, E : pairs of alternatives compared
- optimize over model class \mathcal{M}

$$\min_{X \in \mathcal{M}} \sum_{\alpha, i, j} w_{ij}^{\alpha} (X_{ij} - Y_{ij}^{\alpha})^2$$

- Y_{ij}^{α} measures preference of i over j of voter α . Y^{α} skew-symmetric
- w_{ij}^{α} metric; 1 if α made comparison for $\{i, j\}$, 0 otherwise
- Kemeny optimization:

$$\mathcal{M}_K = \{X \in \mathbb{R}^{n \times n} \mid X_{ij} = \text{sign}(s_j - s_i), s : V \rightarrow \mathbb{R}\}$$

- relaxed version

$$\mathcal{M}_G = \{X \in \mathbb{R}^{n \times n} \mid X_{ij} = s_j - s_i, s : V \rightarrow \mathbb{R}\}$$

- rank-constrained least squares with skew-symmetric matrix variables
- solution is precisely consistent component in HodgeRank

comparisons with other methods

- **analytic hierarchy process (AHP):** take

$$a_{ij} = \begin{cases} \exp(y_{ij}) & \text{if } y_{ij} \text{ exists} \\ 0 & \text{otherwise} \end{cases}$$

A reciprocal matrix, i.e., $a_{ji} = 1/a_{ij} > 0$. Principal eigenvector of A gives global scores [Saaty, 1978].

- **tropical AHP:** principal max-plus eigenvector of Y gives global scores [Elsner–Driessche, 2006]
- suppose $n =$ number of alternatives grows with $m =$ number of voters; when does

$$P(\text{recover top } k \text{ rankings}) \rightarrow 1 \quad \text{as } m, n \rightarrow \infty?$$

Theorem (Tran, 2013)

Under mild assumptions, HodgeRank recovers true ranking of top k items in the above sense. AHP and tropical AHP do not.

Robbins-Monro (1951) algorithm for $A\mathbf{x} = \mathbf{b}$

$$\mathbf{x}_{t+1} = \mathbf{x}_t - \gamma_t(A_t\mathbf{x}_t - \mathbf{b}_t), \quad \mathbb{E}(A_t) = A, \quad \mathbb{E}(\mathbf{b}_t) = \mathbf{b}$$

now consider $\Delta_0\mathbf{s} = \delta_0^* \hat{Y}$, with new rating $Y_t(i_{t+1}, j_{t+1})$

$$\mathbf{s}_{t+1}(i_{t+1}) = \mathbf{s}_t(i_{t+1}) - \gamma_t[\mathbf{s}_t(i_{t+1}) - \mathbf{s}_t(j_{t+1}) - Y_t(i_{t+1}, j_{t+1})]$$

$$\mathbf{s}_{t+1}(j_{t+1}) = \mathbf{s}_t(j_{t+1}) + \gamma_t[\mathbf{s}_t(i_{t+1}) - \mathbf{s}_t(j_{t+1}) - Y_t(i_{t+1}, j_{t+1})]$$

note:

- updates only occur locally on edge $\{i_{t+1}, j_{t+1}\}$
- initial choice: $\mathbf{s}_0 = \mathbf{0}$ or any vector $\sum_i \mathbf{s}_0(i) = 0$
- step size
 - ▶ $\gamma_t = (t + c)^{-\theta}$, $\theta \in (0, 1]$
 - ▶ $\gamma_t = \text{constant}(T)$, e.g. $1/T$ where T is total sample size

averaging process

a second stage averaging process, following \mathbf{s}_{t+1} above

$$\mathbf{z}_{t+1} = \frac{t}{t+1} \mathbf{z}_t + \frac{1}{t+1} \mathbf{s}_{t+1}$$

with $\mathbf{z}_0 = \mathbf{s}_0$

note:

- averaging process speeds up convergence for various choices of γ_t
- one often choose $\gamma_t = c$ to track dynamics
- in this case, \mathbf{z}_t converges to $\hat{\mathbf{s}}$ (population solution), with probability $1 - \delta$, in the (optimal) rate

$$\|\mathbf{z}_t - \hat{\mathbf{s}}\| \leq O\left(t^{-1/2} \cdot \kappa(\Delta_0) \cdot \log^{1/2} \frac{1}{\delta}\right)$$

top Netflix movies according to HodgeRank

Linear Full	Linear 30	Bradley-Terry Full
Greatest Story Ever ...	LOTR III: Return ...	LOTR III: Return ...
Terminator 3	LOTR I: The Fellowship ...	LOTR II: The Two ...
Michael Flatley	LOTR II: The Two ...	LOTR I: The Fellowship ...
Hannibal [Bonus]	Star Wars VI: Return ...	Star Wars V: Empire ...
Donnie Darko [Bonus]	Star Wars V: Empire ...	Raiders of the Lost Arc
Timothy Leary's ...	Star Wars IV: A New Hope	Star Wars IV: A New Hope
In Country	LOTR III: Return ...	Shawshank Redemption
Bad Boys II [Bonus]	Raiders of the Lost Arc	Star Wars VI: Return ...
Cast Away [Bonus]	The Godfather	LOTR III: Return ...
Star Wars: Ewok ...	Saving Private Ryan	The Godfather

- LOTR III shows up twice because of the two DVD editions
- full model has many “bonus” discs that Netflix rents; these are items enjoyed by only a few people

Application to Games

O. Candogan, I. Menache, A. Ozdaglar, and P. Parrilo, “Flows and decompositions of games: harmonic and potential games,” *Math. Oper. Res.*, **36** (2011), no. 3, pp. 474–503.

noncooperative strategic-form finite game

- finite set of **players** $V = \{1, \dots, n\}$
- finite set of **strategies** E_i , for every $i \in V$
- joint strategy space is $E = \prod_{i \in V} E_i$
- **utility function** $u_i : E \rightarrow \mathbb{R}$, $i \in V$
- a game instance is given by the tuple $(V, \{E_i\}_{i \in V}, \{u_i\}_{i \in V})$

strategy

- $\mathbf{p}_i \in E_i$ denotes strategy of player i
- collection of players' strategies is $\mathbf{p} = \{\mathbf{p}_i\}_{i \in V}$, called **strategy profile**
- collection of strategies for all players but the i th one denoted by $\mathbf{p}_{-i} \in E_{-i}$
- $h_i = |E_i|$, cardinality of the strategy space of player i
- $|E| = \prod_{i=1}^n h_i$ for the overall cardinality of the strategy space
- enumerate the actions of players, so that $E_i = \{1, \dots, h_i\}$

Nash equilibrium

- **Nash equilibrium** is strategy profile from which no player can unilaterally deviate and improve its payoff
- formally strategy profile $\mathbf{p} := \{\mathbf{p}_1, \dots, \mathbf{p}_n\}$ is Nash equilibrium if

$$u_i(\mathbf{p}_i, \mathbf{p}_{-i}) \geq u_i(\mathbf{q}_i, \mathbf{p}_{-i}), \quad \text{for every } \mathbf{q}_i \in E_i \text{ and } i \in V$$

- strategy profile $\mathbf{p} := \{\mathbf{p}_1, \dots, \mathbf{p}_n\}$ is ε -equilibrium if

$$u_i(\mathbf{p}_i, \mathbf{p}_{-i}) \geq u_i(\mathbf{q}_i, \mathbf{p}_{-i}) - \varepsilon \quad \text{for every } \mathbf{q}_i \in E_i \text{ and } i \in V$$

- pure Nash equilibrium is an ε -equilibrium with $\varepsilon = 0$

potential game

- a **potential game** is a noncooperative game for which there exists a function $\varphi : E \rightarrow \mathbb{R}$ satisfying

$$\varphi(\mathbf{p}_i, \mathbf{p}_{-i}) - \varphi(\mathbf{q}_i, \mathbf{p}_{-i}) = u_i(\mathbf{p}_i, \mathbf{p}_{-i}) - u_i(\mathbf{q}_i, \mathbf{p}_{-i}),$$

for every $i \in V$, $\mathbf{p}_i, \mathbf{q}_i \in E_i$, $\mathbf{p}_{-i} \in E_{-i}$

- φ is referred to as a **potential** function of the game
- proposed in seminal paper [Monderer–Shapley, 1996]
- widely studied in game theory
- preferences of all players aligned with a global objective
- easy to analyze
- pure Nash equilibrium exists

harmonic games

- Helmholtz decomposition applied to space of **game flows**

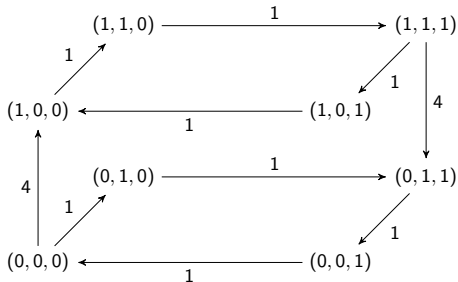
game flow =

potential game \oplus *nonstrategic game* \oplus *harmonic game*

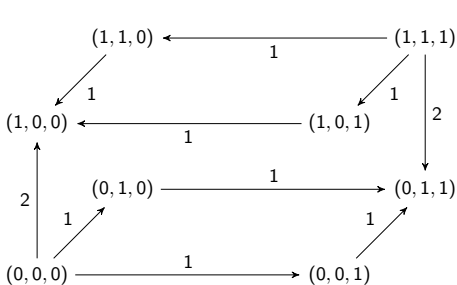
- first defined in [Candogan–Menache–Ozdaglar–Parrilo, 2011] but similar ideas appeared in game theory literature [Hofbauer–Schlag, 2000]
- generically no pure Nash equilibrium
- essentially sums of cycles
- e.g. rock-paper-scissors, cyclic games

example: road sharing game

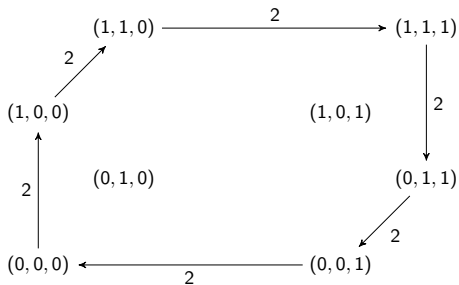
- proposed in [Candogan–Menache–Ozdaglar–Parrilo, 2011]
- three-player game: $V = \{1, 2, 3\}$
- each player choose one of two roads $\{0, 1\}$
- player 3 tries to avoid sharing the road with other players: its payoff decreases by 2 with each other player who shares its road
- player 1 receives a payoff of -1 if player 2 shares its road and 0 otherwise
- payoff of player 2 is equal to negative of the payoff of player 1, i.e., $u_1 + u_2 = 0$
- intuitively, player 1 tries to avoid player 2, whereas player 2 wants to use the same road with player 1



(a) flow of road-sharing game



(b) potential component



(c) harmonic component

Thank You