

Applications of Matroid Methods to Coding Theory

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Outline

1. Correspondence between Matroids and Linear Codes

Application: a matroid-theoretic derivation of the MacWilliams identity

2. Code/Matroid Decomposition

Application: linear-programming (LP) decoding

3. Treewidth of Graphs and Matroids

Application: graphical realizations of codes

Matroids and Codes

Matroids

Definition

A **matroid** is an ordered pair (E, \mathcal{I}) consisting of

- ◇ a finite **ground set** E ; and
- ◇ a collection \mathcal{I} of **independent sets**, which are subsets of E satisfying the following three **independence axioms**:

(I1) $\emptyset \in \mathcal{I}$

(I2) if $I \in \mathcal{I}$, then for any $J \subseteq I$, $J \in \mathcal{I}$

(I3) if $J_1, J_2 \in \mathcal{I}$ with $|J_1| < |J_2|$, then
there exists $e \in J_2 \setminus J_1$ such that $J_1 \cup \{e\} \in \mathcal{I}$

- A subset of E that is not in \mathcal{I} is called a **dependent set**
- A minimal dependent set is called a **circuit**

Vector Matroids

Let

$$A = \begin{bmatrix} | & | & \dots & | \\ \mathbf{v}_1 & \mathbf{v}_2 & \dots & \mathbf{v}_n \\ | & | & & | \end{bmatrix}$$

be a matrix over a field \mathbb{F} .

Take $E = \{1, 2, \dots, n\}$, and define $\mathcal{I} \subseteq 2^E$ via

$I \in \mathcal{I}$ iff the columns \mathbf{v}_i , $i \in I$, are linearly independent over \mathbb{F} .

(E, \mathcal{I}) is a matroid referred to as the **vector matroid** of the matrix A over the field \mathbb{F} ; denoted by $M[A]$ or $M_{\mathbb{F}}[A]$.

A matroid **isomorphic** to some vector matroid over the field \mathbb{F} is said to be **representable** or **\mathbb{F} -representable**.

Graphic Matroids

Let \mathcal{G} be an undirected graph with edge set E .

Define $\mathcal{I} \subseteq 2^E$ via

$I \in \mathcal{I}$ iff I does not contain a cycle of \mathcal{G} .

(E, \mathcal{I}) is a matroid referred to as the **cycle matroid** of the graph \mathcal{G} ; denoted by $M(\mathcal{G})$.

- The circuits of $M(\mathcal{G})$ are precisely the circuits (simple cycles) of \mathcal{G} .

A matroid **isomorphic** to the cycle matroid of some graph is called a **graphic matroid**.

Linear Codes

Let \mathbb{F} be a finite field. An $[n, k]$ linear code over \mathbb{F} is a k -dimensional subspace of \mathbb{F}^n .

We will associate an index set E with a code \mathcal{C} , so that \mathcal{C} is considered to be a subspace of \mathbb{F}^E ; here, $|E| = n$.

A code \mathcal{C} is specified by a generator matrix G , which is a matrix such that $\mathcal{C} = \text{rowspace}_{\mathbb{F}}(G)$;

or equivalently, by a parity-check matrix H , which is a matrix such that $\mathcal{C} = \ker_{\mathbb{F}}(H)$.

The columns of any generator or parity-check matrix of \mathcal{C} are also indexed by the elements of E .

Associating Matroids with Linear Codes

A matrix G over \mathbb{F} determines two different objects:

the vector matroid $M = M_{\mathbb{F}}[G]$;

the code $\mathcal{C} = \text{rowspace}_{\mathbb{F}}(G)$.

Note that if G' is any matrix obtained from G via elementary row operations over \mathbb{F} , then $M_{\mathbb{F}}[G'] = M_{\mathbb{F}}[G]$;

G and G' are just different \mathbb{F} -representations of the same matroid.

Hence, to any linear code \mathcal{C} over \mathbb{F} , we may uniquely assign an \mathbb{F} -representable matroid $M(\mathcal{C})$, by setting $M(\mathcal{C}) := M_{\mathbb{F}}[G]$ for any generator matrix G of \mathcal{C} .

Remark: We could also have set $M(\mathcal{C}) = M_{\mathbb{F}}[H]$ for a *parity-check matrix* H of \mathcal{C} ;
this results in a “dual” version of our exposition.

Aside: MDS Codes

An $[n, k]$ linear code is said to be **maximum distance separable (MDS)** if its minimum distance equals $n - k + 1$.

Fact: If G generates an MDS code of dimension k , then any set of k columns of G is linearly independent; and any set of $k + 1$ columns of G is linearly dependent.

If \mathcal{C} is an $[n, k]$ MDS code, then for the matroid $M(\mathcal{C})$, the collection \mathcal{I} of independent sets is

$$\mathcal{I} = \{I \subseteq [n] : |I| \leq k\}.$$

Such a matroid is called a **uniform matroid**, denoted by $U_{k,n}$.

Bases and Rank: Definitions

$M = (E, \mathcal{I})$ a matroid.

Definition: A **basis** of M is any maximal (wrt inclusion) independent set of M .

By Axiom (I3), all bases of M have the same cardinality.

Definition: The cardinality of any basis of M is called the **rank** of M , denoted by $\text{rank}(M)$ or $r(M)$.

More generally, the **rank function** of M is the function $r : 2^E \rightarrow \mathbb{Z}$ defined as follows: for $X \subseteq E$,

$$r(X) = \max\{|I| : I \in \mathcal{I}, I \subseteq X\}.$$

In particular, $\text{rank}(M) = r(E)$.

Rank and Dimension: Codes

\mathcal{C} a linear code over \mathbb{F} with index set E ;

G a generator matrix for \mathcal{C} ;

$M = M(\mathcal{C}) = M[G]$ the associated matroid.

- Rank function of M : for $X \subseteq E$,

$$r(X) = \text{rank}_{\mathbb{F}}(G|_X) = \dim_{\mathbb{F}}(\mathcal{C}|_X).$$

In particular, $\text{rank}(M) = \text{rank}_{\mathbb{F}}(G) = \dim_{\mathbb{F}}(\mathcal{C})$.

The Dual Matroid

$M = (E, \mathcal{I})$ a matroid, with \mathcal{B} its collection of bases.

For $X \subseteq E$, let $X^c = E - X$.

Define $\mathcal{I}^* = \{I^* : I^* \subseteq B^c \text{ for some } B \in \mathcal{B}\}$.

(E, \mathcal{I}^*) forms a matroid, called the **dual matroid** of M ;
denoted by M^* .

It is clear that M^* has as its collection of bases

$$\mathcal{B}^* = \{B^c : B \in \mathcal{B}\}.$$

Thus, M^* is a matroid on the same ground set as M , but whose bases are the complements of the bases of M .

Duality: Codes

\mathcal{C} a linear code over \mathbb{F} with index set E ;

G a generator matrix for \mathcal{C} ;

$M = M(\mathcal{C}) = M[G]$ the associated matroid.

The **dual code** of \mathcal{C} is defined as

$$\mathcal{C}^\perp = \{\mathbf{x} \in \mathbb{F}^E : \langle \mathbf{c}, \mathbf{x} \rangle_{\mathbb{F}} = 0 \text{ for all } \mathbf{c} \in \mathcal{C}\}.$$

- $M^* = M[H]$ for any parity-check matrix, H , of \mathcal{C} .

Therefore,

$$M^*(\mathcal{C}) \stackrel{\text{def}}{=} (M(\mathcal{C}))^* = M(\mathcal{C}^\perp).$$

In particular, the dual of an \mathbb{F} -representable matroid is also \mathbb{F} -representable.

Deletion and Contraction

Two fundamental operations on a matroid $M = (E, \mathcal{I})$, given an $X \subseteq E$.

Deletion. The matroid $M \setminus X$ is the matroid on ground set $E - X$, whose independent sets are precisely those $I \in \mathcal{I}$ that are contained in $E - X$, *i.e.*,

$$\mathcal{I}(M \setminus X) = \{I \in \mathcal{I} : I \subseteq E - X\}.$$

Contraction. This is the dual operation to deletion:

$$M/X = (M^* \setminus X)^*.$$

Matroid Minors

Definition

A **minor** of a matroid M is any matroid obtained from M via a (possibly empty) sequence of deletion and contraction operations.

Minors are central to matroid theory — e.g., they often turn up in **excluded-minor characterizations**:

- A matroid is binary (*i.e.*, $GF(2)$ -representable) iff it contains no minor isomorphic to the uniform matroid $U_{2,4}$.
- A matroid is regular (*i.e.*, representable over any field) iff it contains no minor isomorphic to any of $U_{2,4}$, $M(\mathcal{H}_7)$ and $M^*(\mathcal{H}_7)$. [Here, \mathcal{H}_7 is the (binary) [7,4] Hamming code.]
- A matroid is graphic iff it contains no minor isomorphic to any of $U_{2,4}$, $M(\mathcal{H}_7)$, $M^*(\mathcal{H}_7)$, $M^*(K_5)$ and $M^*(K_{3,3})$.

Puncturing and Shortening

\mathcal{C} a linear code on index set E , and $X \subseteq E$.

Puncturing. Columns indexed by X deleted from a generator matrix for \mathcal{C} ;

thus, $\mathcal{C} \setminus X$ is the projection of \mathcal{C} onto the coordinates in $E - X$.

Shortening. Columns indexed by X deleted from a parity-check matrix for \mathcal{C} ;

equivalently, \mathcal{C}/X is obtained by taking the subcode of \mathcal{C} that has 0's in all the coordinates in X , and then deleting those coordinates from the subcode.

Then,

$$M(\mathcal{C} \setminus X) = M(\mathcal{C}) \setminus X \quad \text{and} \quad M(\mathcal{C}/X) = M(\mathcal{C})/X.$$

Code Minors

Definition

A **minor** of a code \mathcal{C} is any code obtained from \mathcal{C} via a (possibly empty) sequence of shortening and puncturing operations.

**Application:
The MacWilliams Identity**

The Tutte Polynomial

M a matroid on the ground set E , with rank function r .

Definition: The **Tutte polynomial** of M is defined as

$$T_M(x, y) = \sum_{A \subseteq E} (x - 1)^{r(E) - r(A)} (y - 1)^{|A| - r(A)}$$

Fact: $T_{M^*}(x, y) = T_M(y, x)$.

Remark:

The chromatic polynomial of a graph \mathcal{G} can be obtained as a special case of the Tutte polynomial of $M(\mathcal{G})$.

The MacWilliams Identity

\mathcal{C} an $[n, k]$ linear code over $\mathbb{F} = GF(q)$.

Definition: The **homogeneous weight enumerator polynomial** of \mathcal{C} :

$$W_{\mathcal{C}}(x, y) = \sum_{i=0}^n A_i x^{n-i} y^i,$$

where A_i is the number of codewords of weight i .

Theorem [Greene (1976)]

Let $M = M(\mathcal{C})$. Then,

$$W_{\mathcal{C}}(x, y) = y^{n-k} (x - y)^k T_M \left(\frac{x + (q - 1)y}{x - y}, \frac{x}{y} \right)$$

Corollary (The MacWilliams Identity)

$$W_{\mathcal{C}^\perp}(x, y) = q^{-k} W_{\mathcal{C}}(x + (q - 1)y, x - y)$$

Code Composition/Decomposition

The $\mathcal{S}_m(\mathcal{C}, \mathcal{C}')$ Construction

Let $\mathcal{C}, \mathcal{C}'$ be linear codes of length n, n' , resp., over some field \mathbb{F} ; and let m be an integer s.t. $0 \leq m < \min\{n, n'\}$.

Let $G = [\mathbf{g}_1 \ \mathbf{g}_2 \ \dots \ \mathbf{g}_n]$ and $G' = [\mathbf{g}'_1 \ \mathbf{g}'_2 \ \dots \ \mathbf{g}'_{n'}]$ be generator matrices of \mathcal{C} and \mathcal{C}' , respectively,

Consider the code $\hat{\mathcal{C}}$ with generator matrix

$$\begin{bmatrix} \mathbf{g}_1 & \dots & \mathbf{g}_{n-m} & \mathbf{g}_{n-m+1} & \dots & \mathbf{g}_n & \mathbf{0} & \dots & \mathbf{0} \\ \mathbf{0} & \dots & \mathbf{0} & \mathbf{g}'_1 & \dots & \mathbf{g}'_m & \mathbf{g}'_{m+1} & \dots & \mathbf{g}'_{n'} \end{bmatrix}.$$

Definition

$\mathcal{S}_m(\mathcal{C}, \mathcal{C}')$ is the code of length $n + n' - 2m$ obtained by shortening $\hat{\mathcal{C}}$ at the m “overlapping positions”.

Some Properties of $\mathcal{S}_m(\mathcal{C}, \mathcal{C}')$

Let \mathcal{C}_p and \mathcal{C}_s denote the codes obtained, respectively, by puncturing and shortening \mathcal{C} at its last m coordinates.

Let \mathcal{C}'_p and \mathcal{C}'_s denote the codes obtained, respectively, by puncturing and shortening \mathcal{C}' at its first m coordinates.

Proposition

- (a) $\dim(\mathcal{S}_m(\mathcal{C}, \mathcal{C}')) = \dim(\mathcal{C}) + \dim(\mathcal{C}') - \dim(\mathcal{C}_s \cap \mathcal{C}'_s) - \dim(\mathcal{C}_p + \mathcal{C}'_p).$
- (b) If $\mathcal{C}, \mathcal{C}'$ are codes over a field of characteristic 2, then

$$(\mathcal{S}_m(\mathcal{C}, \mathcal{C}'))^\perp = \mathcal{S}_m(\mathcal{C}^\perp, \mathcal{C}'^\perp).$$

Important Special Cases

$\mathcal{C}, \mathcal{C}'$ linear codes over $\mathbb{F} = GF(q)$;

$$m = (q^{r-1} - 1)/(q - 1), \quad n, n' > 2m.$$

r -sum, $r \geq 1$. $\mathcal{C} \oplus_r \mathcal{C}' = \mathcal{S}_m(\mathcal{C}, \mathcal{C}')$, when

- ◇ $\mathcal{C}_s = \mathcal{C}'_s = \{\mathbf{0}\}$
- ◇ $\mathcal{C}_p = \mathcal{C}'_p = [m, r - 1]$ simplex (*i.e.*, Hamming dual) code

\bar{r} -sum, $r \geq 1$. $\mathcal{C} \bar{\oplus}_r \mathcal{C}' = \mathcal{S}_m(\mathcal{C}, \mathcal{C}')$, when

- ◇ $\mathcal{C}_s = \mathcal{C}'_s = [m, m - (r - 1)]$ Hamming code
- ◇ $\mathcal{C}_p = \mathcal{C}'_p = \{0, 1\}^m$

When $r = 1$, the above definitions degenerate to the **direct sum**:

$$\mathcal{C} \oplus_1 \mathcal{C}' = \mathcal{C} \bar{\oplus}_1 \mathcal{C}' = \mathcal{S}_0(\mathcal{C}, \mathcal{C}') = \mathcal{C} \oplus \mathcal{C}'.$$

Basic Properties of r - and \bar{r} -sums

For the special cases of r - and \bar{r} -sums, the previous proposition specializes to

Corollary

- (a) $\dim(\mathcal{C} \oplus_r \mathcal{C}') = \dim(\mathcal{C}) + \dim(\mathcal{C}') - (r - 1).$
- (b) $\dim(\mathcal{C} \bar{\oplus}_r \mathcal{C}') = \dim(\mathcal{C}) + \dim(\mathcal{C}') - (2^r - r - 1).$
- (c) $(\mathcal{C} \oplus_r \mathcal{C}')^\perp = \mathcal{C}^\perp \bar{\oplus}_r \mathcal{C}'^\perp.$

Remark: For $r = 2$, the definitions of r - and \bar{r} -sum coincide, so that (c) above is in fact

$$(\mathcal{C} \oplus_2 \mathcal{C}')^\perp = \mathcal{C}^\perp \oplus_2 \mathcal{C}'^\perp.$$

**Application:
Linear-Programming (LP) Decoding**

LP Formulation of ML Decoding

Setup:

Binary linear code \mathcal{C} of length n

Discrete memoryless channel: $\Pr[\mathbf{y}|\mathbf{x}] = \prod_{i=1}^n \Pr[y_i|x_i]$

Received word: $\mathbf{y} = (y_1, y_2, \dots, y_n)$

Maximum-Likelihood (ML) Decoding:

determine $\arg \max_{\mathbf{x} \in \mathcal{C}} \Pr[\mathbf{y}|\mathbf{x}]$

Equiv. LP formulation [Feldman, Wainwright, Karger (2005)]:

determine $\arg \min_{\mathbf{x} \in P(\mathcal{C})} \langle \boldsymbol{\gamma}, \mathbf{x} \rangle$, where

$\boldsymbol{\gamma} = (\gamma_1, \dots, \gamma_n)$ with

$$\gamma_i = \log \left(\frac{\Pr[y_i|x_i = 0]}{\Pr[y_i|x_i = 1]} \right)$$

and $P(\mathcal{C}) \stackrel{\text{def}}{=} \text{conv}(\mathcal{C})$ is the **codeword polytope**

Relaxing the LP Formulation

ML decoding is known to be **NP-hard**.

Relax the LP formulation by defining a “looser” set of constraints.

In other words, find “simpler” polytopes $\hat{P}(\mathcal{C}) \subseteq [0, 1]^n$ with $P(\mathcal{C}) \subseteq \hat{P}(\mathcal{C})$, and solve the LP over $\hat{P}(\mathcal{C})$ instead:

$$\arg \min_{\mathbf{x} \in \hat{P}(\mathcal{C})} \langle \gamma, \mathbf{x} \rangle$$

The vertex set of such a polytope $\hat{P}(\mathcal{C})$ contains \mathcal{C} , but also contains extra “**pseudocodeword**” vertices.

Canonical Relaxations

For $H \subseteq \mathcal{C}^\perp$, define

$$Q(H) = \bigcap_{\mathbf{h} \in H} P(\mathbf{h}^\perp)$$

where $\mathbf{h}^\perp = \{\mathbf{x} \in \{0, 1\}^n : \langle \mathbf{h}, \mathbf{x} \rangle \equiv 0 \pmod{2}\}$.

LP Decoding: determine $\arg \min_{\mathbf{x} \in Q(H)} \langle \gamma, \mathbf{x} \rangle$

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Question: For which codes \mathcal{C} do there exist $H \subseteq \mathcal{C}^\perp$
such that $Q(H)$ has no pseudocodewords?

Answer: **Geometrically perfect codes**,
i.e., codes \mathcal{C} such that $P(\mathcal{C}) = Q(\mathcal{C}^\perp)$
(codeword polytope = full canonical relaxation).

Interlude — Cycle Codes of Graphs

Given a graph $\mathcal{G} = (V, E)$, the **cycle code** of \mathcal{G} is the *binary* linear code whose *parity-check matrix* is the $|V| \times |E|$ vertex-edge incidence matrix of \mathcal{G} .

We will denote the cycle code of \mathcal{G} by $\mathcal{C}[\mathcal{G}]$.

Note: $M(\mathcal{C}[\mathcal{G}]) = M^*(\mathcal{G})$.

A Characterization of Geom. Perfect Codes

An excluded-minor characterization ...

Theorem

[Barahona and Grötschel (1986), based on Seymour (1982)]

A binary linear code \mathcal{C} is geometrically perfect iff \mathcal{C} does not contain as a minor any code equivalent to one of the following:

- ◇ the $[7,3]$ Hamming dual, \mathcal{H}_7^\perp ;
- ◇ a certain $[10,5]$ isodual code, R_{10} ; and
- ◇ the dual of the cycle code of K_5 , i.e., $\mathcal{C}[K_5]^\perp$.

An Alternative Characterization

A characterization via code decompositions ...

Theorem

[Grötschel and Truemper (1989), based on Seymour (1982)]

A binary linear code \mathcal{C} is geometrically perfect iff \mathcal{C} can be constructed by means of coordinate permutations, direct-sums, 2-sums and 3-sums starting with codes, each of which is **a minor of \mathcal{C}** , and each of which is one of the following:

- ◇ **the cycle code of some graph;**
- ◇ **the $[7, 4]$ Hamming code;**
- ◇ **$\mathcal{C}(K_{3,3})^\perp$;**
- ◇ **$\mathcal{C}(V_8)^\perp$.**

Corollaries of the Decomposition Theorem

Let \mathcal{G} be the family of geometrically perfect codes.

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- Therefore, there is a polynomial-time maximum-likelihood decoding algorithm for codes in \mathcal{G} .

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- Therefore, there is a polynomial-time maximum-likelihood decoding algorithm for codes in \mathcal{G} .
- \mathcal{G} is **not asymptotically good**: codes from \mathcal{G} cannot have both min. dist. and dimension growing linearly with codelength.
- Therefore, pseudocodewords cannot be avoided when LP decoding is applied to good codes.

Tree Decompositions of Graphs and Matroids

Tree Decompositions of Graphs

Let \mathcal{G} be a graph with vertex set $V(\mathcal{G})$.

A **tree decomposition** of \mathcal{G} consists of a **tree** T , and an ordered collection $\mathcal{V} = (V_x, x \in V(T))$ of subsets of $V(\mathcal{G})$, satisfying

- $\bigcup_{x \in V(T)} V_x = V$;
- for each $v \in V(\mathcal{G})$, the subgraph of T induced by $\{x \in V(T) : v \in V_x\}$ is **connected**; and
- for each pair of **adjacent vertices** $u, v \in V(\mathcal{G})$, we have $\{u, v\} \subseteq V_x$ for some $x \in V(T)$.

We then define **width** $(T, \mathcal{V}) \stackrel{\text{def}}{=} \max_{x \in V(T)} |V_x| - 1$.

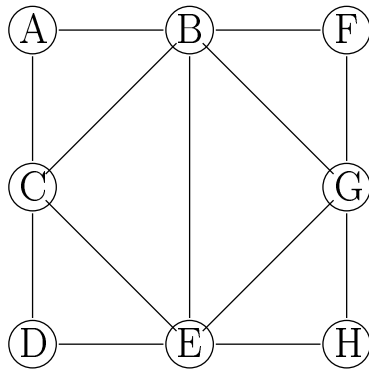
Treewidth of Graphs

Definition [Robertson & Seymour (1983)]

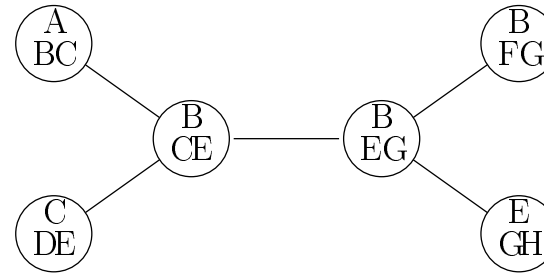
The **treewidth** of \mathcal{G} is defined to be the least width of any tree decomposition of \mathcal{G} ; denoted by $\kappa_{\text{tree}}(\mathcal{G})$.

Some Examples

- For any tree T , $\kappa_{\text{tree}}(T) = 1$.
- If \mathcal{G} is a cycle on at least three vertices, then $\kappa_{\text{tree}}(\mathcal{G}) = 2$.
- The graph \mathcal{G} shown below also has treewidth 2.



\mathcal{G}



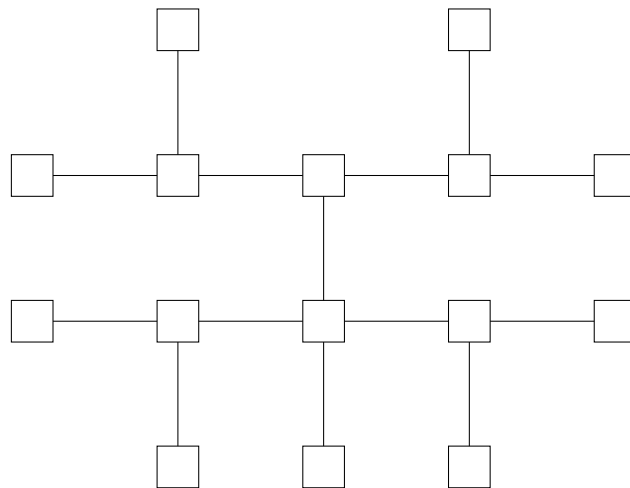
An optimal tree decomposition of \mathcal{G}

Tree Decompositions of Matroids

M a matroid on ground set E , with rank function r .

A **tree decomposition** of M is a pair (T, ω) , where

- T is a tree, and

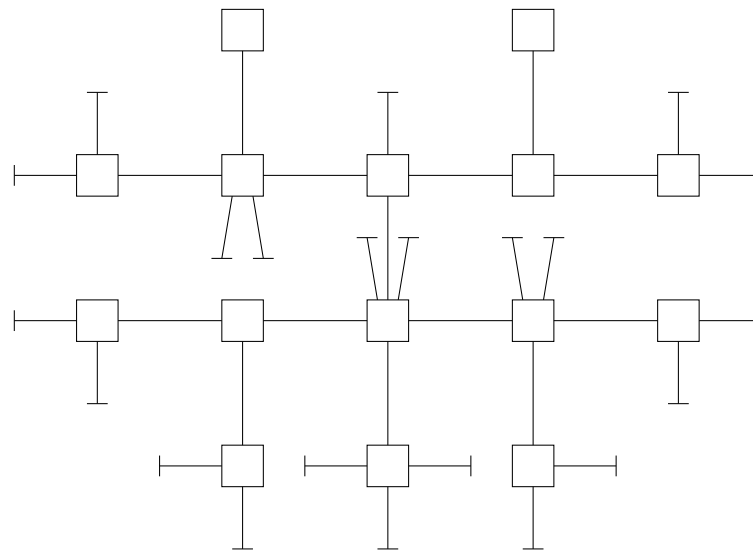


Tree Decompositions

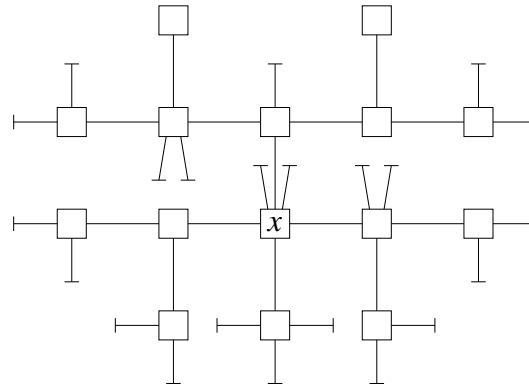
M a matroid on ground set E , with rank function r .

A **tree decomposition** of M is a pair (T, ω) , where

- T is a tree, and
- $\omega : E \rightarrow V(T)$ is a mapping.

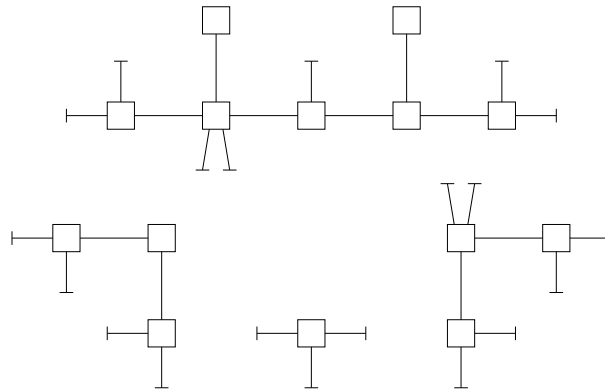


Node-width in a tree decomposition



Given a tree decomposition (T, ω) of M , and a node $x \in V(T)$

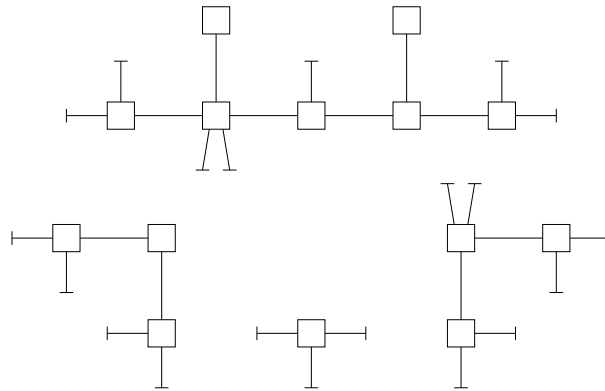
Node-width in a tree decomposition



Given a tree decomposition (T, ω) of M , and a node $x \in V(T)$:

- the removal of x from T yields a disconnected graph whose components, T_1, \dots, T_δ , are subtrees of T

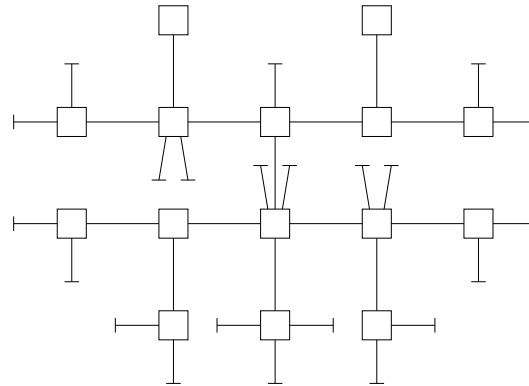
Node-width in a tree decomposition



Given a tree decomposition (T, ω) of M , and a node $x \in V(T)$:

- the removal of x from T yields a disconnected graph whose components, T_1, \dots, T_δ , are subtrees of T
- for $j = 1, \dots, \delta$, set $F_j = \omega^{-1}(V(T_j))$
- $\text{node-width}(x) = \sum_{i=1}^{\delta} r(E - F_i) - (\delta - 1) \text{rank}(M)$

Matroid Treewidth



$$\text{width}(T, \omega) = \max_{x \in V(T)} \text{node-width}(x)$$

Definition [Hliněný and Whittle (2006);
attributed to Jim Geelen]:

The **treewidth** of M is defined to be

$$\kappa_{\text{tree}}(M) = \min_{(T, \omega)} \text{width}(T, \omega).$$

Relating Graph and Matroid Treewidth

Theorem [Hliněný and Whittle (2006)] For any graph \mathcal{G} ,

$$\kappa_{\text{tree}}(M(\mathcal{G})) = \kappa_{\text{tree}}(\mathcal{G}).$$

It is known that the problem of computing the treewidth of a graph is NP-hard, and therefore, so is the corresponding problem for matroids.

**Application:
Graphical Models of Codes**

Graphical Models of Codes

Graphical models of codes and the associated message-passing decoding algorithms are a major focus area of modern coding theory.

Graphical models come in many flavours:

- **Trellises** (the Viterbi decoding algorithm)
- **Tanner graphs**
- **Factor graphs**
- **Normal graphical models/realizations** [Forney (2001)]

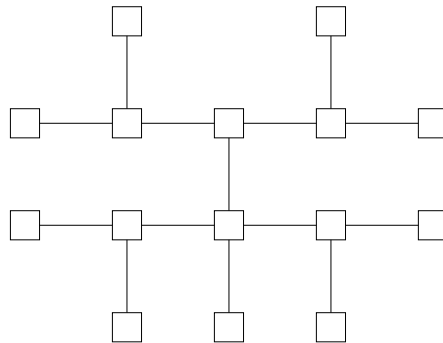
The decoding algorithms commonly associated with these models are variants of the abstract **Generalized Distributive Law**, as expounded by **Aji & McEliece (2000)**.

Graph Decompositions

Let \mathcal{C} be a linear code defined on an index set I .

A **graph decomposition** of (the index set of) \mathcal{C} is a pair (\mathcal{G}, ω) , where

- \mathcal{G} is a connected graph, and

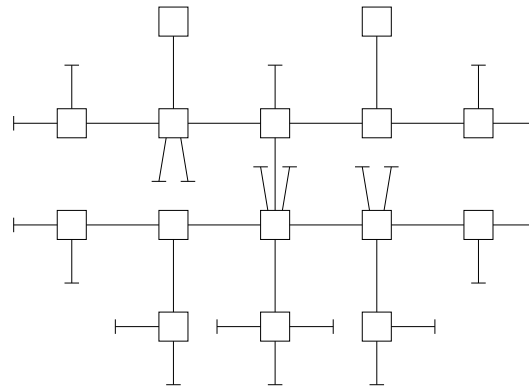


Graph Decompositions

Let \mathcal{C} be a linear code defined on an index set I .

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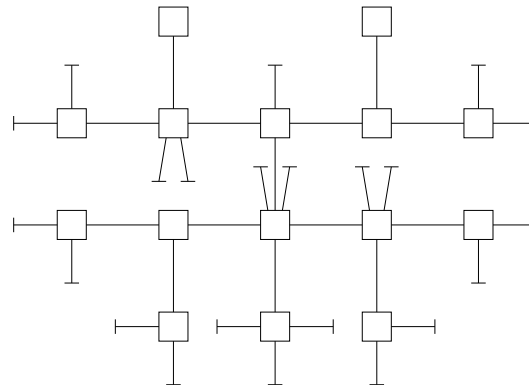


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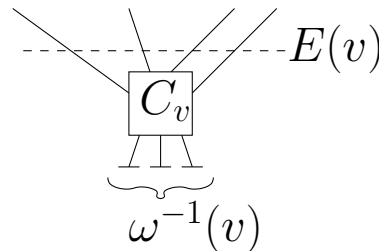
When \mathcal{G} is a tree, (\mathcal{G}, ω) is called a **tree decomposition**.

Normal Graphical Models

For a graph $\mathcal{G} = (V, E)$, given $v \in V$, let $E(v)$ denote the set of edges of \mathcal{G} incident with v .

A graph decomposition (\mathcal{G}, ω) of a code \mathcal{C} can be **extended** to a **normal graphical model** $(\mathcal{G}, \omega, (\mathcal{S}_e, e \in E), (C_v, v \in V))$, where

- for each $e \in E$, \mathcal{S}_e is a vector space over \mathbb{F} , called a **state space**;
- for each $v \in V$, C_v is a subspace of $\mathbb{F}^{\omega^{-1}(v)} \oplus \left(\bigoplus_{e \in E(v)} \mathcal{S}_e \right)$, called a **local constraint (code)**.



(Normal) Graphical Realizations

A **valid global configuration** of a normal graphical model Γ is a vector of the form $\mathbf{b} = ((x_i, i \in I), (\mathbf{s}_e, e \in E))$, where

- for each $i \in I$, x_i is a symbol from \mathbb{F} ;
- for each $e \in E$, \mathbf{s}_e is a state from \mathcal{S}_e ;
- for each $v \in V$, $((x_i, i \in \omega^{-1}(v)), (\mathbf{s}_e, e \in E(v))) \in C_v$.

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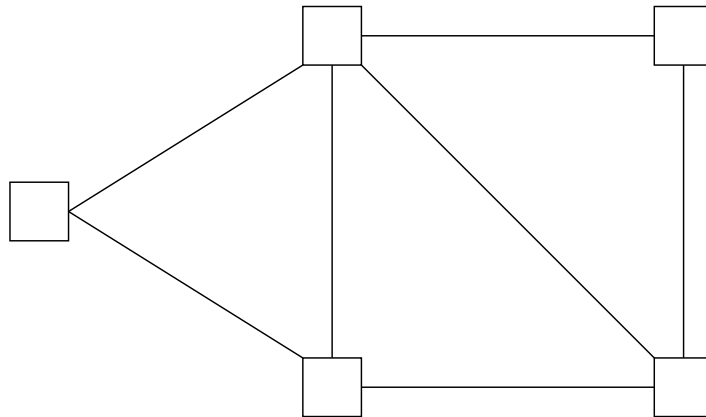
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If $\mathfrak{B}|_I = \mathcal{C}$, then Γ is called a **(normal) graphical realization** of \mathcal{C} .

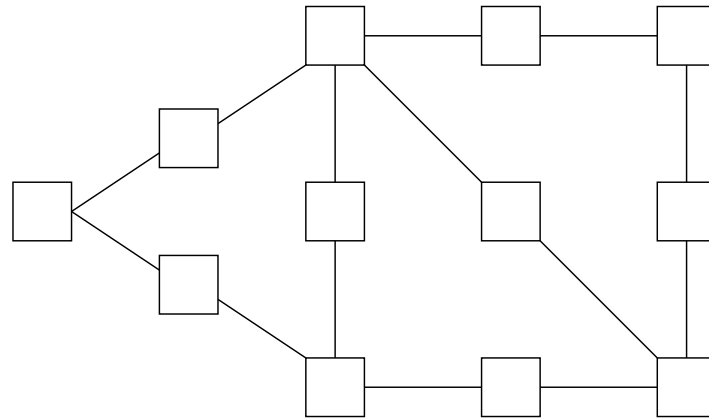
An Example

Consider an arbitrary graph \mathcal{G}_0 :



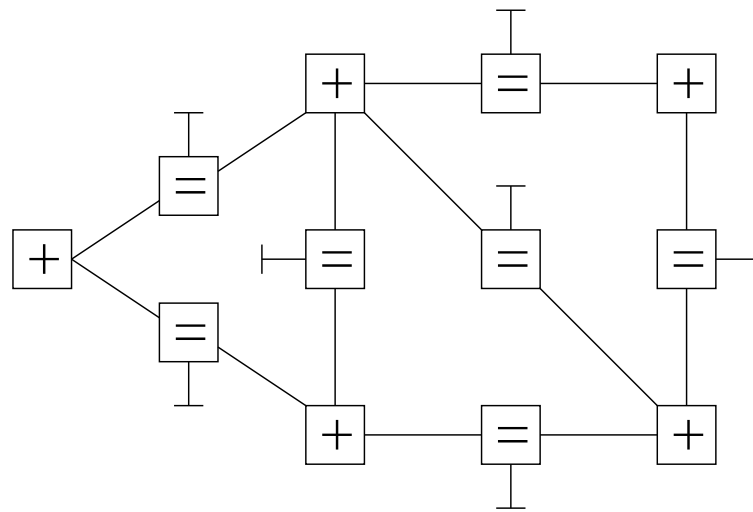
An Example

Subdivide the edges of \mathcal{G}_0 to form \mathcal{G} :



An Example

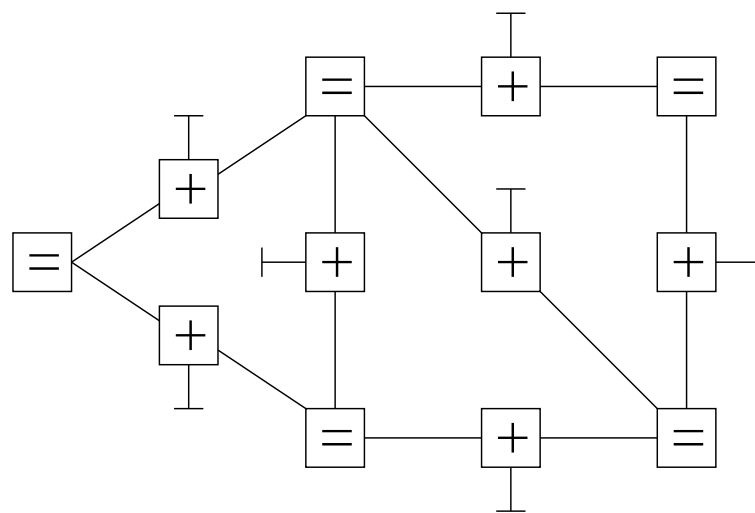
Construct a graphical model (over \mathbb{F}_2) on \mathcal{G} as depicted below:



This is a graphical realization of the **cycle code** $\mathcal{C}[\mathcal{G}_0]$.

The Dual Example

Replace all $+$'s by $=$'s, and vice versa:



This is a graphical realization of the **dual of $\mathcal{C}[\mathcal{G}_0]$** . [Forney (2001)]

Constraint Complexity of a Realization

Any graphical realization of code has a natural associated decoding algorithm, namely, the **sum-product algorithm** [Forney (2001)].

The computational complexity of the sum-product algorithm is determined in large part by the dimensions of the local constraint codes in the realization.

Definition: Let $\Gamma = (\mathcal{G}, \omega, (C_v, v \in V), (\mathcal{S}_e, e \in E))$ be a graphical realization of a code \mathcal{C} . The **constraint complexity** of Γ is defined to be

$$\kappa(\Gamma) = \max_{v \in V} \dim(C_v).$$

How Low Can You Go?

Given: a code \mathcal{C} and a connected graph \mathcal{G}

Fact: Any graph decomposition (\mathcal{G}, ω) of (the index set of) \mathcal{C} can be extended to a graphical realization of \mathcal{C} .

Question: How small can the constraint complexity of a realization of \mathcal{C} on \mathcal{G} be?

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Let $\mathfrak{R}(\mathcal{C}; \mathcal{G}, \omega)$ denote the set of all realizations of \mathcal{C} that extend a given graph decomposition (\mathcal{G}, ω) .

$$\kappa(\mathcal{C}; \mathcal{G}, \omega) = \min_{\Gamma \in \mathfrak{R}(\mathcal{C}; \mathcal{G}, \omega)} \kappa(\Gamma)$$

$$\kappa(\mathcal{C}; \mathcal{G}) = \min_{\omega} \kappa(\mathcal{C}; \mathcal{G}, \omega)$$

Tree Realizations

A tree realization of a code \mathcal{C} is a graphical realization of \mathcal{C} in which the underlying graph is a tree.

Since the realization is cycle-free, the associated sum-product algorithm gives an exact implementation of **maximum-likelihood (ML) decoding** [Forney (2001)], [Aji & McEliece (2001)].

Minimal Tree Realizations

For a given **tree decomposition** (T, ω) of a code \mathcal{C} , **Forney (2001)** gave a canonical method of constructing a tree realization in $\mathfrak{R}(\mathcal{C}; T, \omega)$.

Forney's construction can be shown to minimize, among all realizations in $\mathfrak{R}(\mathcal{C}; T, \omega)$, the dimension of the local constraint at each vertex of T [**K. (2007)**].

Let $\mathcal{M}(\mathcal{C}; T, \omega)$ denote this **minimal tree realization**; thus

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Forney (2003) gave an explicit expression for the dimensions of the local constraints in $\mathcal{M}(\mathcal{C}; T, \omega)$.

Treewidth of Codes

Definition

Treewidth: $\kappa_{\text{tree}}(\mathcal{C}) = \min_{(T, \omega)} \kappa(\mathcal{C}; T, \omega)$
(minimum over all tree decompositions of \mathcal{C})

Fact:

Forney's expression for the dimensions of the local constraints in a minimal tree realization shows that

$$\kappa_{\text{tree}}(\mathcal{C}) = \kappa_{\text{tree}}(M(\mathcal{C}))$$

Thus, $\kappa_{\text{tree}}(M(\mathcal{C}))$ may be viewed as a measure of the ML-decoding complexity of \mathcal{C} .

Realizations on Graphs with Cycles

There is little known about the problem of finding low-complexity realizations of a code \mathcal{C} on a given connected graph \mathcal{G} , when \mathcal{G} is not a tree.

When \mathcal{G} is a simple cycle, the problem is one of finding optimal **tailbiting trellis realizations** of codes, which has been studied by Koetter and Vardy (2003).

Halford and Chugg (2008) gave a lower bound on $\kappa(\mathcal{C}; \mathcal{G})$ in terms of “forest-inducing edge cuts” of \mathcal{G} .

Their lower bound is subsumed by (a slight modification of) the following bound [K. (2009)]:

$$\kappa(\mathcal{C}; \mathcal{G}) \geq \frac{\kappa_{\text{tree}}(\mathcal{C})}{\kappa_{\text{tree}}(\mathcal{G}) + 1}$$

Other Complexity Measures

One can define the **pathwidth** of graphs, matroids, and codes, by considering only those tree decompositions in which the underlying tree is a **simple path**.

These notions are related to each other much like treewidth.

The **pathwidth** of a linear code is essentially the same as its **minimal trellis complexity**.

Some Interesting Results

The connections between pathwidth and treewidth of graphs, matroids, and codes can be exploited to show that

- computing the minimal trellis complexity (among all coordinate permutations) of a code is NP-hard [K. (2008)]
- the ratio between the pathwidth and the treewidth of a code grows at most logarithmically with codelength, and a logarithmic rate of growth is in fact achievable [K. (2009)]
- “Good” families of codes cannot have realizations of bounded complexity on graphs of bounded treewidth [K. (2009)]

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