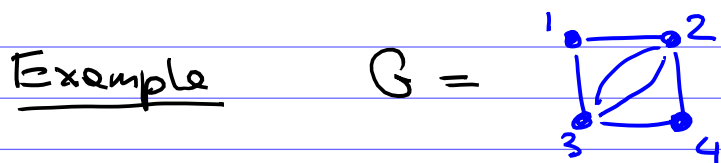


## Kirchhoff's Matrix Tree Theorem (MTT)

$G$  - a graph on  $n$  vertices



$L = (l_{ij})$  - the Laplacian matrix of  $G$   
 $n \times n$  matrix

$$l_{ij} = \begin{cases} - \# \text{ edges between } i \text{ \& } j & \text{if } i \neq j \\ \text{deg}_G(i) & \text{if } i = j \end{cases}$$

Reduced Laplacian matrix:

$\tilde{L} = L$  with  $n^{\text{th}}$  row &  $n^{\text{th}}$  column removed  
 $(n-1) \times (n-1)$  matrix

Instead of  $n$ , we can pick any other  $i = 1, \dots, n$

Example

$$L = \begin{bmatrix} 2 & -1 & -1 & 0 \\ -1 & 4 & -2 & -1 \\ -1 & -2 & 4 & -1 \\ 0 & -1 & -1 & 2 \end{bmatrix}, \quad \tilde{L} = \begin{bmatrix} 2 & -1 & -1 & 0 \\ -1 & 4 & -2 & -1 \\ -1 & -2 & 4 & -1 \\ \del{0} & \del{-1} & \del{-1} & \del{2} \end{bmatrix}$$

$4 \times 4$   $3 \times 3$

Matrix Tree Theorem.

$$\# \text{ Spanning trees of } G = \det \tilde{L}.$$

## Incidence Matrix of $G$

should not be confused with the adjacency matrix

$n \times m$  matrix

$n = \#$  vertices of  $G$

$m = \#$  edges of  $G$

The  $(i, j)$  entry of the incidence matrix is 1 if  $i^{\text{th}}$  vertex is incident to the  $j^{\text{th}}$  edge. Otherwise, it is 0.

---

We'll need a modification of the incidence matrix:

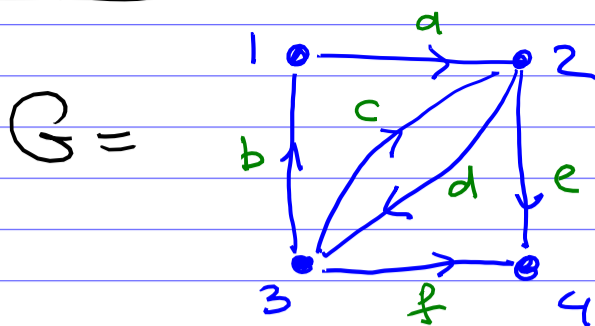
## Oriented Incidence Matrix

Orient all edges of  $G$ .

$B = (b_{ij})$   $n \times m$  matrix

$b_{ij} =$   $\left\{ \begin{array}{l} 1 \text{ if the } i^{\text{th}} \text{ vertex is the source of the } j^{\text{th}} \text{ edge} \\ -1 \text{ if the } i^{\text{th}} \text{ vertex is the target of the } j^{\text{th}} \text{ edge} \\ 0 \text{ otherwise} \end{array} \right.$

## Example



$$B = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ -1 & 0 & -1 & 1 & 1 & 0 \\ 0 & 1 & 1 & -1 & 0 & -1 \\ 0 & 0 & 0 & 0 & -1 & 1 \end{bmatrix} \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix}$$

$a \quad b \quad c \quad d \quad e \quad f$

Lemma  $L = B \cdot B^T$

Proof  $(B \cdot B^T)_{ij}$  is the

dot-product of the  $i^{\text{th}}$  and  $j^{\text{th}}$  rows of  $B$ .

$i=j$ : #  $\pm 1$ 's in the  $i^{\text{th}}$  row of  $B = \deg_G(i)$ .

$$\text{So } (B B^T)_{ii} = \deg_G(i).$$

$i \neq j$ : For any edge between vertices  $i$  &  $j$ , the corresp. entries of the  $i^{\text{th}}$  &  $j^{\text{th}}$  rows are  $1$  and  $-1$ , or  $-1$  and  $1$ . For any other edge, at least one of these entries is  $0$ .

So  $(B B^T)_{ij} = -\# \text{ edges between vertices } i \text{ \& } j$ .

□

## Reduced Oriented Incidence Matrix

$\tilde{B} = B$  with the  $n^{\text{th}}$  row removed.

$(n-1) \times m$  matrix.

### Example

$$\tilde{B} = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ -1 & 0 & -1 & 1 & 1 & 0 \\ 0 & 1 & 1 & -1 & 0 & -1 \\ \del{0 & 0 & 0 & 0 & -1 & 1} \end{bmatrix}$$

$3 \times 6$  matrix

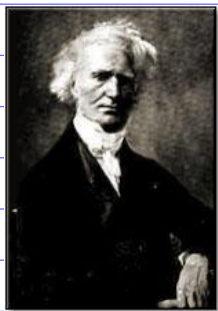
Lemma  $L = \tilde{B} \cdot \tilde{B}^T$

Proof. The same proof.  $\square$

## Cauchy - Binet Formula



Baron Augustin-Louis Cauchy  
(1789-1857)



Jacques Philippe Marie Binet  
(1786-1856)

Let  $k \leq m$ .  $C$  -  $k \times m$  matrix,

$D$  -  $m \times k$  matrix

$$\det(C \cdot D) = \sum_{\substack{S \subseteq \{1, 2, \dots, m\} \\ |S|=k}} \det(C_S) \cdot \det(D_S)$$

where  $C_S$  and  $D_S$  are square  $k \times k$  submatrices of  $C$  and  $D$ , respectively, located in column set  $S$ .

## Example

$$C = \begin{bmatrix} 1 & 2 & 3 \\ 2 & -1 & -1 \end{bmatrix}$$

$$D = \begin{bmatrix} 3 & 5 \\ 1 & 1 \\ -2 & 3 \end{bmatrix}$$

$$C \cdot D = \begin{bmatrix} -1 & 16 \\ 7 & 6 \end{bmatrix}$$

$$\det(C \cdot D) = \begin{vmatrix} 1 & 2 \\ 2 & -1 \end{vmatrix} \cdot \begin{vmatrix} 3 & 5 \\ 1 & 1 \end{vmatrix}$$

$$+ \begin{vmatrix} 1 & 3 \\ 2 & -1 \end{vmatrix} \cdot \begin{vmatrix} 3 & 5 \\ -2 & 3 \end{vmatrix} + \begin{vmatrix} 2 & 3 \\ -1 & -1 \end{vmatrix} \cdot \begin{vmatrix} 1 & 1 \\ -2 & 3 \end{vmatrix}$$

$$= (-5) \cdot (-2) + (-7) \cdot 19 + 1 \cdot 5$$

$$= 10 - 133 + 5 = -118.$$

$$\begin{vmatrix} -1 & 16 \\ 7 & 6 \end{vmatrix} = -118.$$

## Proof of Cauchy - Binet

Consider the product of square  $(k+m) \times (k+m)$  matrices:

$$\begin{matrix} k \\ m \end{matrix} \begin{bmatrix} I_k & C \\ 0 & I_m \end{bmatrix} \cdot \begin{bmatrix} C & 0 \\ -I_m & D \end{bmatrix} \begin{matrix} k \\ m \end{matrix}$$

$$= \begin{matrix} k \\ m \end{matrix} \begin{bmatrix} 0 & C \cdot D \\ -I_m & D \end{bmatrix} \begin{matrix} k \\ m \end{matrix}$$

Here  $I_k$  is  $k \times k$  identity matrix  
 $0$  is the zero matrix

$$\det \begin{bmatrix} I_k & C \\ 0 & I_m \end{bmatrix} = 1$$

$$\det \begin{bmatrix} 0 & C \cdot D \\ -I_m & D \end{bmatrix} = \pm \det(C \cdot D)$$

(more precisely,  $= (-1)^{m \cdot (k+1)} \det(C \cdot D)$ ).

$$\det \begin{bmatrix} C & 0 \\ -I_m & D \end{bmatrix} = \pm \text{R.H.S of Cauchy - Binet formula.}$$

(Check that the signs agree!)

Example ( $C, D$  from prev. example)

$$\det \begin{bmatrix} C & 0 \\ -I_m & D \end{bmatrix} =$$

$$= \det \left[ \begin{array}{ccc|cc} 1 & 2 & 3 & 0 & 0 \\ 2 & -1 & -1 & 0 & 0 \\ \hline -1 & 0 & 0 & 3 & 5 \\ 0 & -1 & 0 & 1 & 1 \\ 0 & 0 & -1 & -2 & 3 \end{array} \right]$$

expand this using the definition of  $\det$  as an alternating sum over permutations...

# back to Matrix Tree Theorem

## Proof of MTT

reduced oriented incidence matrix

$$\det(\tilde{L}) = \det(\tilde{B} \cdot \tilde{B}^T)$$

Cauchy Binet

$$\sum_{S \subseteq \{1, \dots, m\}} \left( \det \tilde{B}_S \right)^2$$

$$S \subseteq \{1, \dots, m\}$$

$$|S| = n-1$$

sum over all  $(n-1)$ -element subsets of edges

$n = \# \text{ vertices}$   
 $m = \# \text{ edges}$

## Lemma

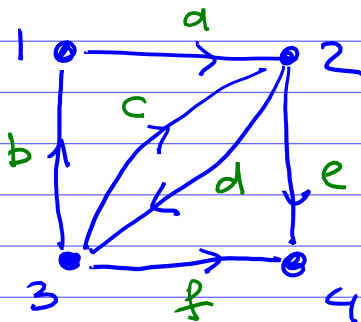
$$\det(\tilde{B}_S) = \pm 1 \quad \text{if}$$

$S$  is the set of edges of a spanning tree of  $G$

Otherwise,  $\det(\tilde{B}_S) = 0$ .

Example

$G =$



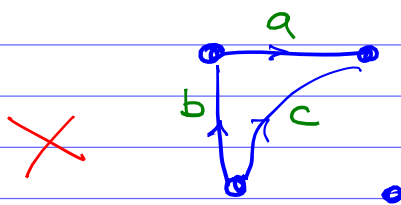
$$\tilde{B} = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ -1 & 0 & -1 & 1 & 1 & 0 \\ 0 & 1 & 1 & -1 & 0 & -1 \\ \hline 0 & 0 & 0 & 0 & -1 & 1 \end{bmatrix}$$

a   b   c   d   e   f

$$\det(\tilde{B}_{\{a,b,c\}}) = \begin{vmatrix} 1 & -1 & 0 \\ -1 & 0 & -1 \\ 0 & 1 & 1 \end{vmatrix} = 0$$

why? ↙

a   b   c

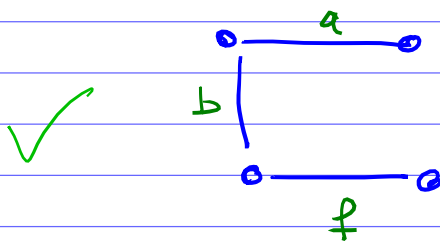


not a spanning tree

$$\det(\tilde{B}_{\{a,b,f\}}) = \begin{vmatrix} 1 & -1 & 0 \\ -1 & 0 & 0 \\ 0 & 1 & -1 \end{vmatrix} =$$

a   b   c

$$= 1.$$



a spanning tree



Clearly, the last lemma implies  $\det \tilde{L} = \det (\tilde{B} \cdot \tilde{B}^T) = \# \text{ spanning trees of } G$ .

It remains to prove the lemma.

### Proof of Lemma

1. If  $S$  is not the set of edges of a spanning tree.

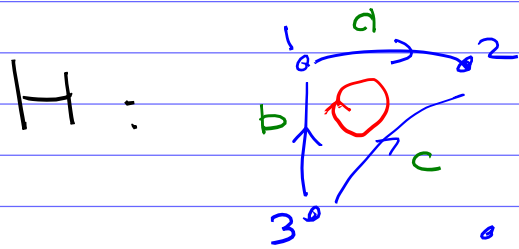
$S$  forms a subgraph  $H$  of  $G$  with  $n-1$  edges, which is not a tree

$\Rightarrow$  The subgraph contains a cycle  $C$ .

$\Rightarrow$  the columns of  $\tilde{B}$  corresponding to edges of  $C$  are linearly dependent

$$\Rightarrow \det (\tilde{B}_S) = 0$$

Example  $S = \{a, b, c\}$



$$\tilde{B} = \begin{bmatrix} 1 & -1 & 0 & \dots & \dots \\ -1 & 0 & -1 & \dots & \dots \\ 0 & 1 & 1 & \dots & \dots \end{bmatrix}$$

a      b      c

$$+ \text{Column}_a + \text{Column}_b - \text{Column}_c = \vec{0}$$

↑ " + " if the orient. of edge agrees with the orient. of cycle  
 ↑ " - " if the orient. of edge is opposite to the orient. of cycle

2. If  $S$  is the set of edges of a spanning tree  $T$  of  $G$ .

Induction on  $n$ .

Base  $n=1$ . (Easy to check)

Inductive step  $n \geq 2$ .

Find a leaf  $l$  of  $T$ ,  
s.t.  $l \neq n$ . (Any tree on  $\geq 2$  vertices has  $\geq 2$  leaves.)

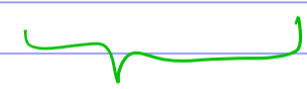
The row of  $\tilde{B}_S$  corresp. to  $l$  contains exactly one non-zero entry, which is equal to  $\pm 1$ .

Expand  $\det(\tilde{B}_S)$  by  $l^{\text{th}}$  row. We get

$$\det(\tilde{B}_S) = \pm \det(\tilde{\tilde{B}})$$

similar (but smaller) matrix corresponding to the tree obtained from  $T$  by removing the leaf  $l$ .

$$= \dots = \pm 1.$$



keep removing leaves one by one.

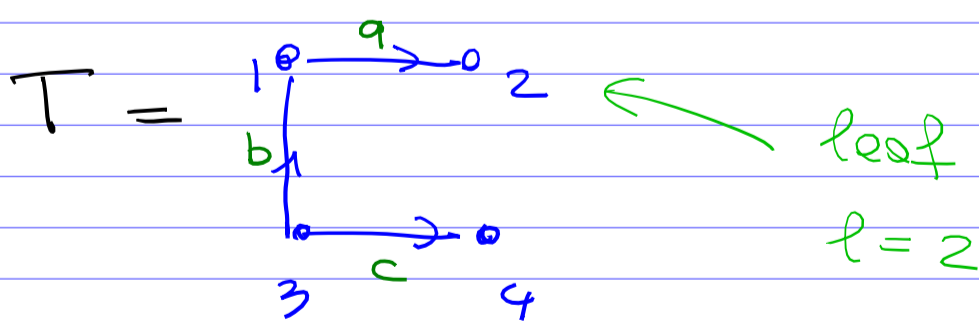
Example  $S = \{a, b, f\}$

$$\tilde{B} = \begin{bmatrix} 1 & -1 & 0 \\ -1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

$a$ 
 $b$ 
 $f$

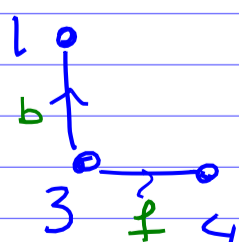
$$\tilde{B}_{\{a, b, f\}} = \begin{bmatrix} 1 & -1 & 0 \\ -1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

row 2



$$\begin{vmatrix} 1 & -1 & 0 \\ -1 & 0 & 0 \\ 0 & 1 & 1 \end{vmatrix} = \pm \begin{vmatrix} -1 & 0 \\ 1 & 1 \end{vmatrix} = \pm |1|$$

$= \pm 1.$



□

This finishes the proof of the MTT.

Example. Let's deduce Cayley's formula from the MTT.

$$G = K_n.$$

$$L = \begin{bmatrix} n-1 & -1 & \dots & -1 \\ -1 & n-1 & & \\ \vdots & & \ddots & \\ -1 & -1 & \dots & n-1 \end{bmatrix} \quad \begin{array}{l} n \times n \\ \text{matrix} \end{array}$$

$$\tilde{L} = \begin{bmatrix} n-1 & -1 & \dots & -1 \\ -1 & n-1 & & \\ \vdots & & \ddots & \\ -1 & -1 & \dots & n-1 \end{bmatrix} \quad \begin{array}{l} (n-1) \times (n-1) \\ \text{matrix.} \end{array}$$

How to calculate  $\det(\tilde{L})$ ?

Let's find all eigenvalues of  $\tilde{L}$ .

$$\tilde{L} - nI_{n-1} = \begin{bmatrix} -1 & -1 & \dots & -1 \\ -1 & -1 & \dots & -1 \\ \dots & \dots & \dots & \dots \\ -1 & -1 & \dots & -1 \end{bmatrix}$$

$(n-1) \times (n-1)$  matrix of rank 1.

So its all eigenvalues, except one, equal to zero.

Also  $\sum \text{eigenvalues} = \text{trace}$

Thus the eigenvalues of

$$\tilde{L} - nI_{n-1} \text{ are}$$

$$\underbrace{0, 0, \dots, 0}_{n-2}, -n+1$$

$$\uparrow \\ \text{tr}(\tilde{L} - nI)$$

$\Rightarrow$  the eigenvalues of  $\tilde{L}$  are

$$\underbrace{n, n, \dots, n}_{n-2}, 1$$

$\Rightarrow \det \tilde{L} = \prod \text{eigenvalues}$

$$= \underbrace{n \cdot n \cdot \dots \cdot n}_{n-2} \cdot 1 = n^{n-2}.$$

# Hypercubes

Let's now calculate #  
spanning trees of the  
 $d$ -hypercube graph  $H_d$ .

$H_d$  = the 1-skeleton of  
the  $d$ -dimensional  
hypercube.

Examples

$$H_1 = \bullet \text{---} \bullet$$

1 spanning  
tree

$$H_2 = \begin{array}{c} \bullet \text{---} \bullet \\ | \quad | \\ \bullet \text{---} \bullet \end{array}$$

4 spanning  
trees

$$H_3 = \begin{array}{c} \bullet \text{---} \bullet \text{---} \bullet \\ | \quad | \quad | \\ \bullet \text{---} \bullet \text{---} \bullet \\ | \quad | \quad | \\ \bullet \text{---} \bullet \text{---} \bullet \end{array}$$

?  
spanning  
trees.

$$H_4 = \begin{array}{c} \bullet \text{---} \bullet \text{---} \bullet \text{---} \bullet \\ | \quad | \quad | \quad | \\ \bullet \text{---} \bullet \text{---} \bullet \text{---} \bullet \\ | \quad | \quad | \quad | \\ \bullet \text{---} \bullet \text{---} \bullet \text{---} \bullet \\ | \quad | \quad | \quad | \\ \bullet \text{---} \bullet \text{---} \bullet \text{---} \bullet \end{array}$$

# Product of graphs

$$G = (V_1, E_1), \quad H = (V_2, E_2)$$

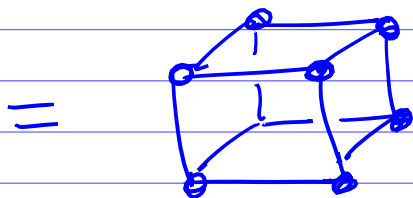
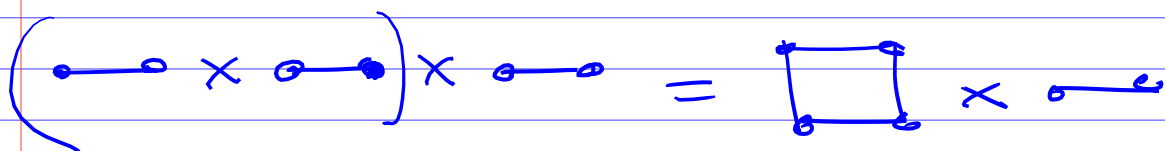
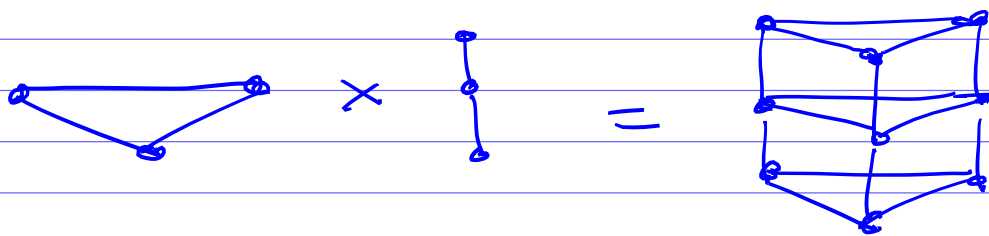
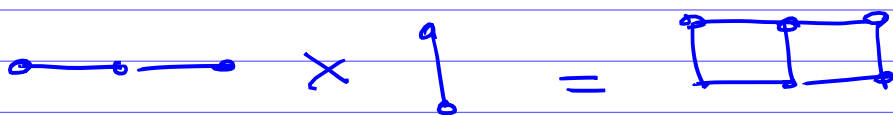
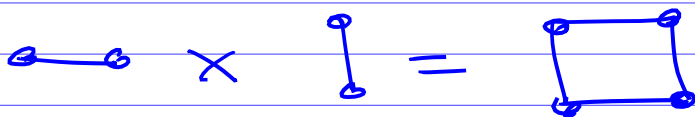
$$G \times H := (V, E)$$

where  $V = V_1 \times V_2$

$$:= \{(u, v) \mid u \in V_1, v \in V_2\}$$

$$E = \left\{ ((u, v), (u', v')) \mid \begin{array}{l} u, u' \in V_1; v, v' \in V_2 \\ u = u' \text{ and } (v, v') \in E_2, \text{ or} \\ v = v' \text{ and } (u, u') \in E_1 \end{array} \right\}$$

## Examples



In general, the hypercube graph is the product:

$$H_d = \underbrace{\bullet \text{---} \bullet \times \bullet \text{---} \bullet \times \dots \times \bullet \text{---} \bullet}_d$$

---

Let  $A(G)$  be the adjacency matrix of  $G$ .

Lemma. Suppose that

$A(G)$  has eigenvalues

$\alpha_1, \dots, \alpha_m$ ; and

$A(H)$  has eigenvalues

$\beta_1, \beta_2, \dots, \beta_n$

Then  $A(G \times H)$  has  $m \cdot n$

eigenvalues  $\alpha_i + \beta_j$ ,

$i \in [m], j \in [n]$ .

Proof idea. Use Linear Algebra.

This is related to tensor products of matrices

$$A(G \times H) = A(G) \otimes I_n + I_m \otimes A(H).$$

Exercise. Prove this lemma.

Example

$$G = H = \begin{array}{c} \bullet \text{---} \bullet \\ 1 \quad 2 \end{array}$$

$$A(G) = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

eigenvalues are  $1, -1$

$$A(G \times H) = A \left( \begin{array}{cc} \bullet & \bullet \\ \bullet & \bullet \end{array} \right)$$

$$\begin{array}{c} 11 \\ 12 \\ e1 \\ 22 \end{array} \left[ \begin{array}{cc|cc} 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ \hline 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{array} \right] \begin{array}{c} 11 \\ 12 \\ 21 \\ 22 \end{array}$$

has eigenvalues

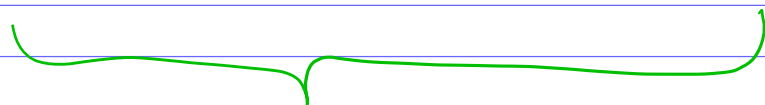
$$1+1, 1-1, -1+1, -1-1$$

$$= 2, 0, 0, -2$$

In general  $A(H_d)$  has

eigenvalues

$$\pm 1 \pm 1 \pm 1 \dots \pm 1$$



$d$  terms

$2^d$  choices for

signs



We've got

Lemma.  $A(H_d)$  has

eigenvalues  $-d + 2k$  with  
multiplicity  $\binom{d}{k}$  for  $k=0, \dots, d$

---

Example  $A(H_3)$  has

eigenvalues

$$\begin{array}{cccc} -3, & -1, & -1, & -1, & 1, & 1, & 1, & 3 \\ \underbrace{\hspace{1.5cm}} & \underbrace{\hspace{3.5cm}} & \underbrace{\hspace{3.5cm}} & \underbrace{\hspace{1.5cm}} & & & & \\ \binom{3}{0}=1 & \binom{3}{1}=3 & \binom{3}{2}=3 & \binom{3}{3}=1 & & & & \end{array}$$

How does it help us?

We need the (reduced)  
Laplacian matrix, not the  
adjacency matrix?

Observation  $H_d$  is a

$d$ -regular graph.

(all vertices have the same  
degree  $d$ ).

$$S_0 \quad L(H_d) = dI - A(H_d)$$

Its eigenvalues are  
 $d$  - the eigenvalues of  $A(H_d)$ .

Corollary  $L(H_d)$  has the eigenvalues:

$2k$  with multiplicity  $\binom{d}{k}$   
for  $k = 0, 1, \dots, d$ .

Theorem (follows from MTT).

Suppose  $L(G)$  has eigenvalues  $\lambda_1 = 0, \lambda_2, \dots, \lambda_n$

Then # spanning trees of  $G$

$$= \det(\tilde{L}) = \frac{\lambda_2 \cdot \lambda_3 \cdot \dots \cdot \lambda_n}{n}$$

(where  $n$  is the number of vertices of  $G$ ).

---

Exercise. Let  $L$  be any symmetric  $n \times n$  matrix with all row and column sums equal to zero. Let  $\tilde{L}$  be the  $(n-1) \times (n-1)$  matrix obtained from  $L$  by removing the last row & the last column.

Let  $\lambda_1 = 0, \lambda_2, \dots, \lambda_n$  be the eigenvalues of  $L$ .

$$\text{Then } \det(\tilde{L}) = \frac{\lambda_2 \cdot \lambda_3 \cdot \dots \cdot \lambda_n}{n}.$$

Now we deduce.

Theorem. The hypercube graph  $H_d$  has

$$\frac{1}{2^d} \prod_{k=1}^d (2k)^{\binom{d}{k}}$$

$$= \boxed{2^{2^d - d - 1} \prod_{k=1}^d k^{\binom{d}{k}}}$$

Spanning trees.