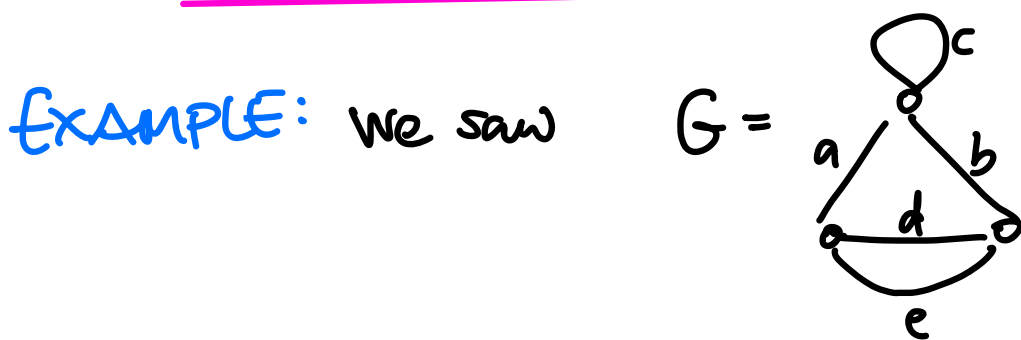


# Counting spanning trees (§2.4) and directed Euler tours (not in book)

Recall that we already defined for a multigraph  $G = (V, E)$ , a **spanning tree** for  $G$  is a subset  $T \subset E$  for which  $(V, T)$  is a tree.



has **5** spanning trees:  $\left\{ \begin{array}{c} T_1 \\ \text{a} \text{---} \text{b} \\ \text{a} \end{array} \right\} \left\{ \begin{array}{c} T_2 \\ \text{a} \text{---} \text{d} \\ \text{a} \end{array} \right\} \left\{ \begin{array}{c} T_3 \\ \text{d} \text{---} \text{b} \\ \text{d} \end{array} \right\} \left\{ \begin{array}{c} T_4 \\ \text{a} \text{---} \text{e} \\ \text{a} \end{array} \right\} \left\{ \begin{array}{c} T_5 \\ \text{b} \text{---} \text{e} \\ \text{b} \end{array} \right\}$

We'll learn how to count these, and some slightly fancier things.

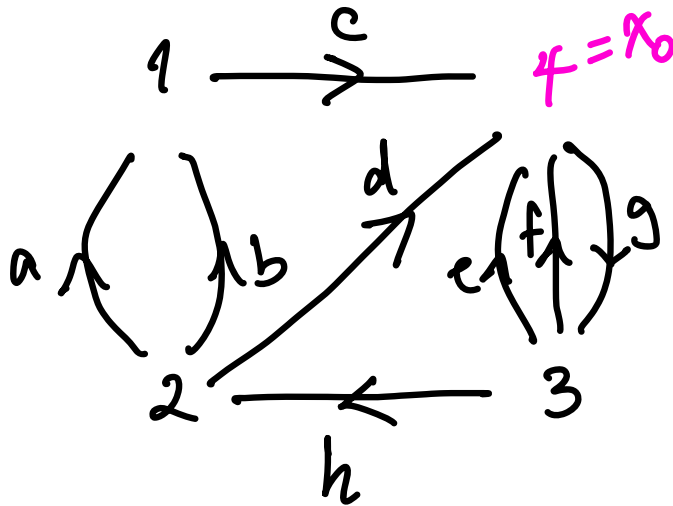
## DEFINITION:

In a digraph  $D = (V, A)$  with  $x_0 \in V$ ,

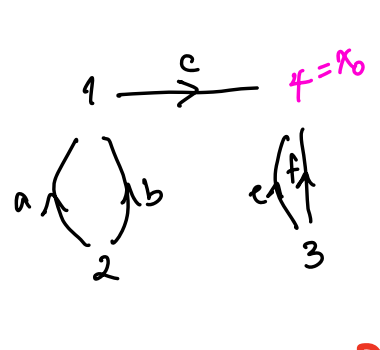
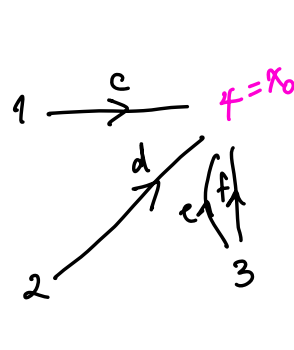
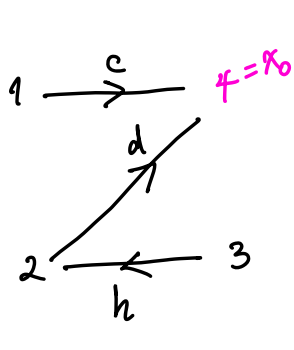
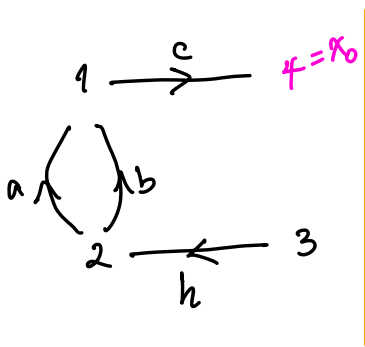
a **spanning tree directed toward  $x_0$**  is a subset  $T \subset A$  whose underlying undirected graph  $G = (V, T)$  is a tree, and every  $y \in V$  has a **unique directed path**  $y \rightarrow \dots \rightarrow x_0$  to  $x_0$  in  $D$ .

# EXAMPLE

$\mathcal{D} =$



has 9 spanning trees directed toward  $x_0 = 4$ :



$\{ ach, bch, cdh, cde, cdf, ace, acf, bce, bcf \}$

We'll compute these tree enumerators.

## DEFINITION:

For  $G = (V, E)$  an undirected multigraph,

$$t(G) := \# \{ \text{spanning trees } T \text{ for } G \}$$

e.g.  $t(\text{graph}) = 5$

$$t(G; \underline{e}) := \sum_{\text{spanning trees } T \text{ for } G} \prod_{e \in T} e$$

e.g.  $t(\text{graph}; a, b, c, d, e)$

$$= ab + ad + bd + ae + be$$

a list of variables, one for each edge in  $E$



But for any  $G=(V,E)$ , one can also compute

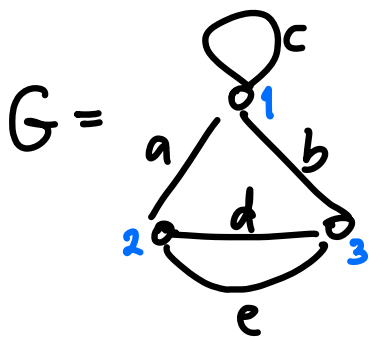
$t(G; \underline{e})$  from  $t(\vec{G}, x_0; \underline{a})$  by building

$\vec{G}=(V,A)$  with one pair of antiparallel arcs  $E, E'$  for each (non-loop) undirected edge  $e \in E$ ,

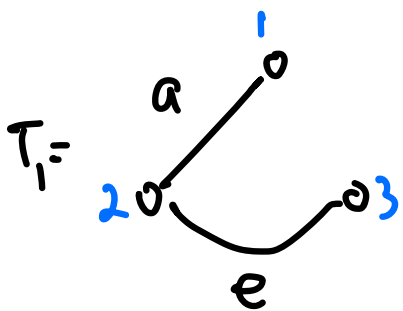
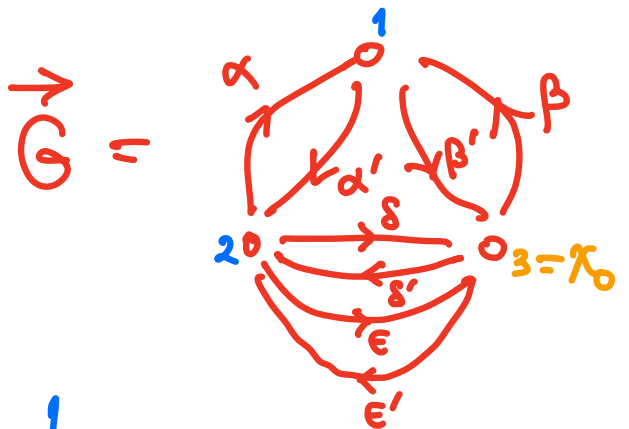
$$\text{and then } t(G, \underline{e}) = \left[ t(\vec{G}, x_0; \underline{a}) \right]_{\substack{E=e \\ E'=e}}$$

(regardless of the choice of  $x_0 \in V$ ).

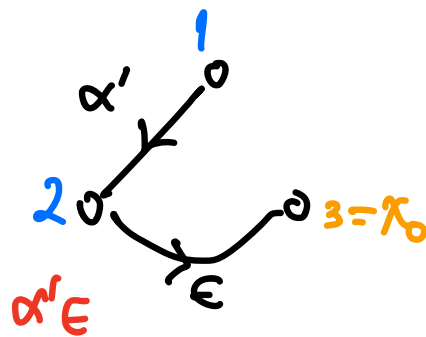
### EXAMPLE:



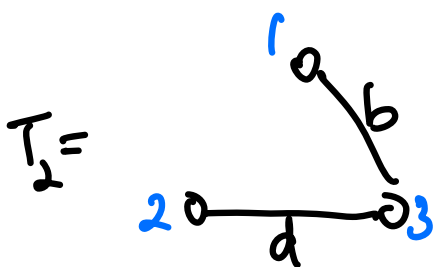
$\rightsquigarrow$



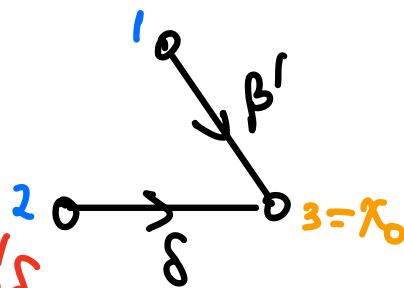
$\longleftrightarrow$



$ae \leftarrow \alpha' \epsilon$



$\longleftrightarrow$



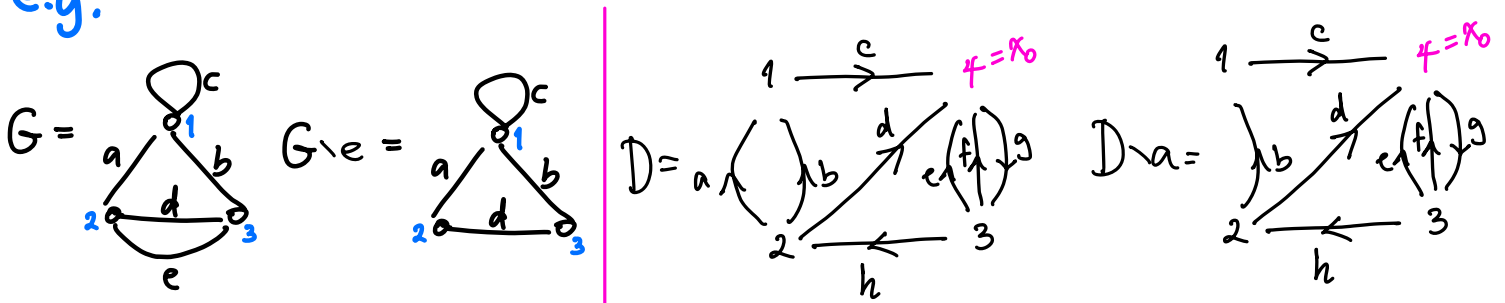
$bd \leftarrow \beta' \delta$

One compute  $t(G)$ ,  $t(G; e)$ ,  $t(D, x_0; a)$  via certain **recursions** (although **not** very computationally efficient, taking  $2^N$  steps if  $N = |E|, |A|$ ), using two fundamental operations.

**DEFINITION:** Given  $G = (V, E)$  (multigraph) or  $D = (V, A)$  (digraph) and  $e \in E$  or  $a \in A$ ,

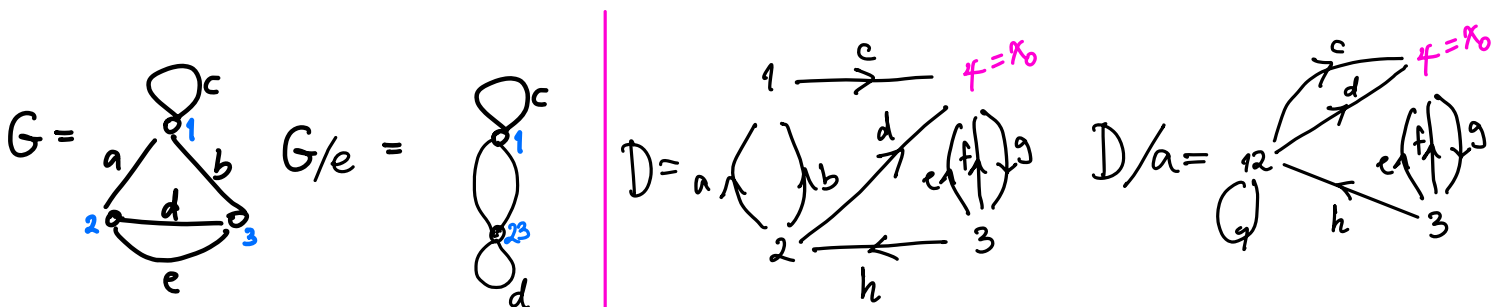
- the **deletion**  $G \setminus e := (V, E - \{e\})$   
 $D \setminus a := (V, A - \{a\})$

e.g.



and **non-loop**  $e \in \bar{E}$  or  $a \in \bar{A}$ , so  $x \neq y$ ,  
 $\{x, y\}$   $\{x, y\}$

- the **contraction**  $G / e := (V / x=y, E - \{e\})$   
 $D / a := (V / x=y, A - \{a\})$



**PROPOSITION:** One can compute  $t(G)$ ,  $t(G; \underline{e})$ ,  $t(D, x_0; \underline{a})$  via these recursions and initial conditions:

$$(a) \quad 1 = t(\square_G) = t(\square_G; \underline{e}) = t(\square_D, x_0; \underline{a})$$

(b) If  $\hat{G} = G$  with all loops removed  
 $\hat{D} = D$  with all loops removed,

then  $t(G) = t(\hat{G})$ ,  $t(G; \underline{e}) = t(\hat{G}; \underline{e})$   
 $t(D; \underline{a}) = t(\hat{D}; \underline{a})$

(c)  $0 = t(G) = t(G; \underline{e})$  whenever  $G$  is disconnected,  
 $0 = t(D, x_0; \underline{a})$  whenever  $\exists y \in V$  with  
 no directed path  $y \rightarrow \dots \rightarrow x_0$  in  $D$ .

(d) If  $e \in E$  is non-loop, then

$$\begin{cases} t(G) = t(G - e) + t(G/e) \\ t(G; \underline{e}) = t(G - e, \underline{e}) + e \cdot t(G/e) \end{cases}$$

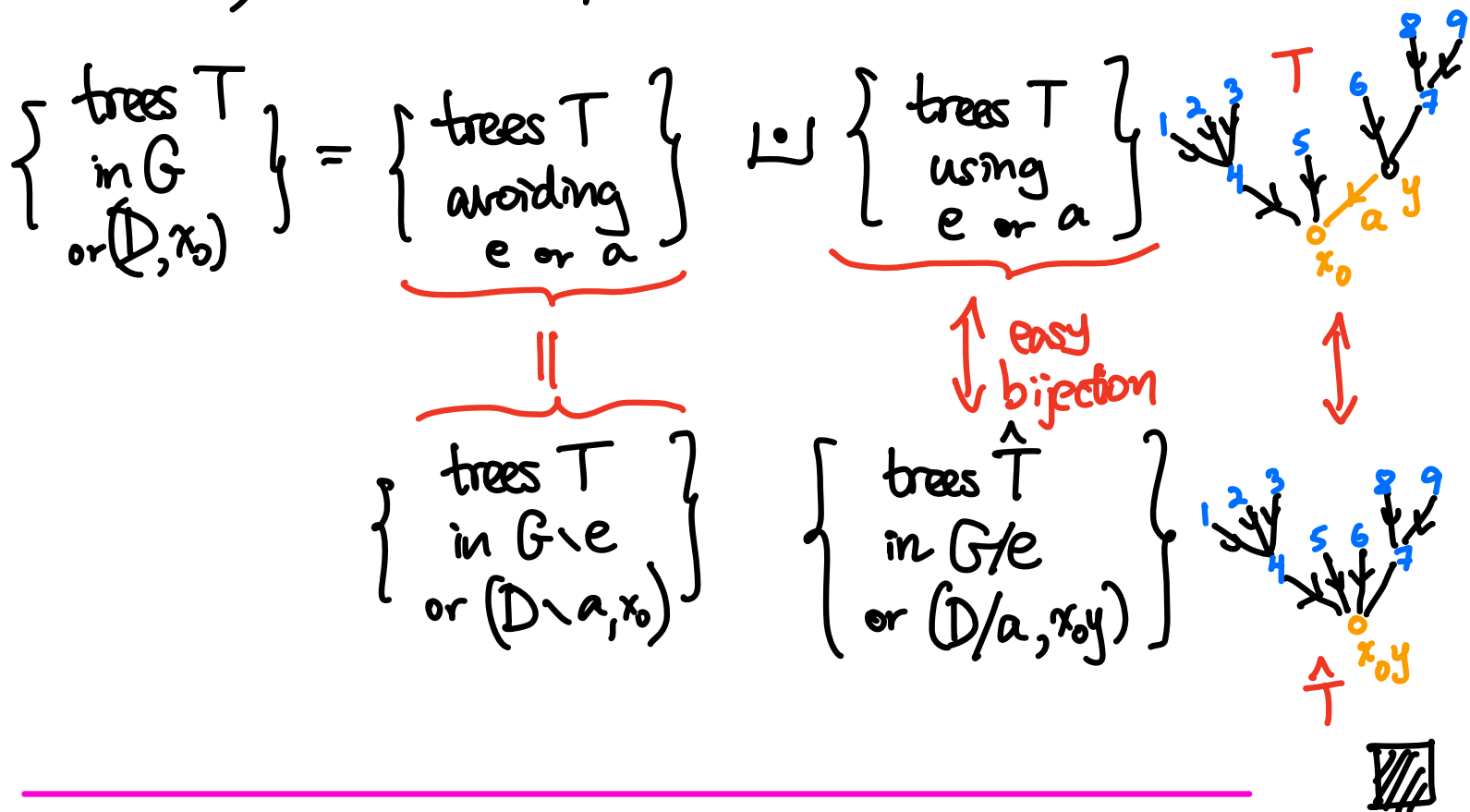
DELETION-  
CONTRACTION  
RECURSIONS

If  $a \in A$  is non-loop and points toward  $x_0$ , then  
 $y \xrightarrow{a} x_0$

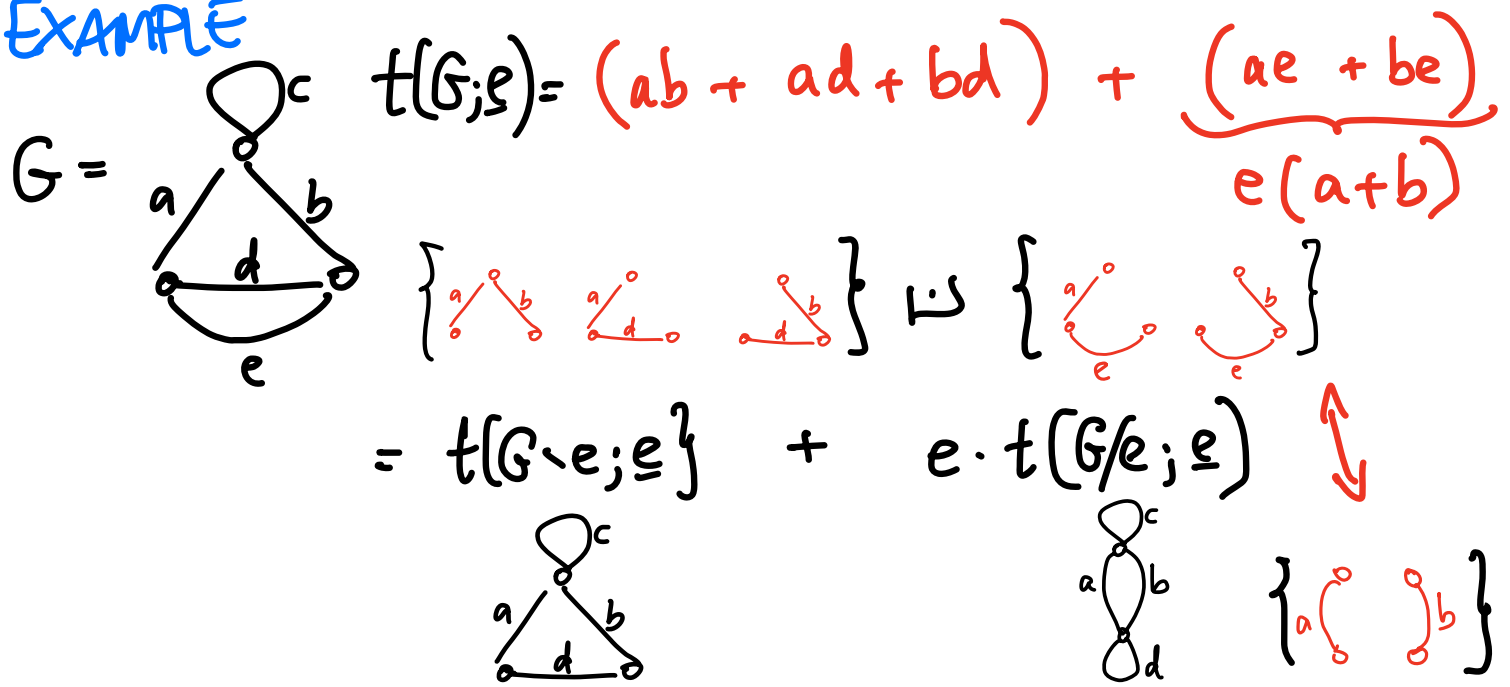
$$\hookrightarrow \begin{cases} t(D, x_0; \underline{a}) = t(D \setminus a, x_0; \underline{a}) + a \cdot t(D/a, x_0; \underline{a}) \end{cases}$$

proof: (a), (b), (c) are all pretty straightforward.

For (d), one decomposes the set of trees  $T$  as follows:

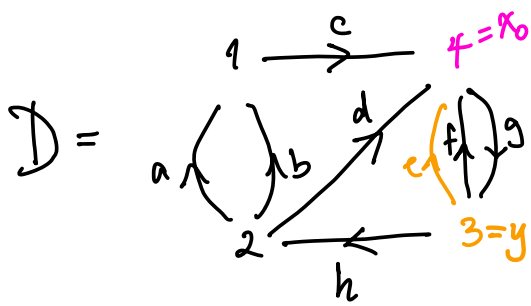


**EXAMPLE**



**ACTIVE LEARNING**

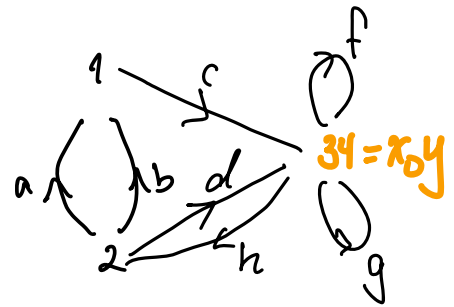
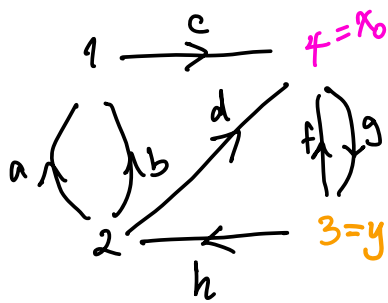
Compute  $t \left( \begin{array}{c} \text{graph with vertices at top and bottom} \\ \text{edges: } a, b, c, d, e \end{array} ; \underline{e} \right)$  via deletion-contraction.



$$\{ach, bch, cdh, cdf, acf, bcf\} \sqcup \{cde, ace, bce\}$$

$$t(D, x_0; \underline{a}) = (ach + bch + cdh + cdf + acf + bcf) + \frac{(cde + ace + bce)}{e(cd + ac + bc)}$$

$$= t(D \setminus e, x_0; \underline{a}) + e \cdot t(D/e, x_0 y; \underline{a})$$



As an algorithm for computing  $t(G)$ ,  $t(G; \underline{e})$ ,  $t(D, x_0; \underline{a})$ , deletion-contraction takes  $2^{|E|}$  or  $2^{|A|}$  steps, so it bogs down quickly. But it helps us prove a faster method using **Laplacian matrices**

**DEFINITION:** Given  $G = (V, E)$  or  $D = (V, A)$  the  $|V| \times |V|$  **Laplacian matrices**  $L(G)$ ,  $L(D)$  are defined by

$$L(G)_{x,y} \text{ or } L(D)_{x,y} = \begin{cases} \sum_{\text{non-loops arcs/edges } a \text{ out of } x} a & \text{if } x=y \\ -\sum_{\text{arcs/edges } x \xrightarrow{a} y} a & \text{if } x \neq y \end{cases}$$



## EXAMPLES:

$$L(G) = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} a+b & -a & -b \\ -a & a+d+e & -(d+e) \\ -b & -(d+e) & b+d+e \end{bmatrix} \end{matrix}$$

$$L(D) = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} c & 0 & 0 & -c \\ -(a+b) & a+b+d & 0 & -d \\ 0 & -h & e+fh & -(e+f) \\ 0 & 0 & -g & g \end{bmatrix} \end{matrix}$$

## THEOREM (Kirchhoff's Matrix-Tree Theorem)

For any multigraph  $G=(V,E)$  or digraph  $D=(V,A)$   
and any vertex  $x_0 \in V$ ,

$$t(G; e) = \det L(G)^{x_0, x_0}$$

where  $L := L - \begin{cases} \text{row } x_0, \\ \text{column } x_0 \end{cases}$

$$t(D, x_0; a) = \det L(D)^{x_0, x_0}$$

## EXAMPLE:

$$L(G) = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} a+b & -a & -b \\ -a & a+d+e & -(d+e) \\ -b & -(d+e) & b+d+e \end{bmatrix} \end{matrix}$$

$x_0 = 3$

$$\begin{aligned} \det L(G)^{x_0, x_0} &= \det \begin{bmatrix} a+b & -a \\ -a & a+d+e \end{bmatrix} \\ &= (a+b)(a+d+e) - a^2 \\ &= \cancel{a^2} + ad + ae + ab + bd + be - \cancel{a^2} \end{aligned}$$

$$= ad + ae + ab + bd + be$$

$$L(D) = \begin{array}{c} 1 \\ 2 \\ 3 \\ x_0=4 \end{array} \left[ \begin{array}{ccc|c} c & 0 & 0 & -c \\ -(a+b) & a+b+d & 0 & -d \\ 0 & -h & e+f+h & -(e+f) \\ \hline 0 & 0 & -g & g \end{array} \right] \begin{array}{l} y=x_0 \\ \\ \\ \end{array}$$

$$\det L(D)^{x_0, x_0} = \det \begin{bmatrix} c & 0 & 0 \\ -(a+b) & a+b+d & 0 \\ 0 & -h & e+f+h \end{bmatrix} = c(a+b+d)(e+f+h)$$

### ACTIVE LEARNING:

Explain why  $\det L(D) = 0 = \det L(G)$  always.

### proof of Kirchhoff's Matrix-Tree Theorem:

We'll do the proof for  $D=(V, A)$ , and then it follows for  $G=(V, E)$  via the construction  $G \mapsto \vec{G}$  from before.

The proof will show  $t(D, x_0; \underline{a}) = \det(L(D)^{x_0, x_0})$

by induction on  $|A|$ , by checking that  $\det(L(D)^{x_0, x_0})$  satisfies all the initial conditions (a), (b), (c) and the deletion-contraction recursion (d) that computes  $t(D, x_0; \underline{a})$ .

For (a),  $1 = t(\square, x_0; \underline{a}) = \det \begin{pmatrix} 1 \end{pmatrix}$  is correct.

For (b),  $t(\hat{D}, x_0; \underline{a}) = t(D, x_0; \underline{a})$  if  $\hat{D} = D$  with loops removed, this is consistent since  $L(D) = L(\hat{D})$  (it ignores loops).

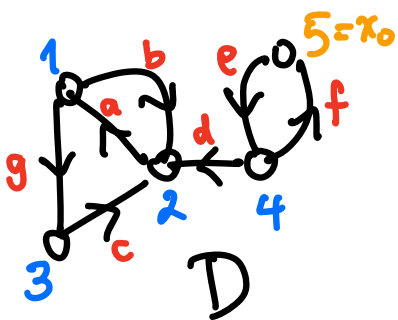
For (c), if  $\exists y \in V$  with no directed path  $y \rightarrow \dots \rightarrow x_0$ , consider the set  $V'$  of all such  $y$ , and the square submatrix  $L'$  of  $L(D)$  indexed by rows, columns in  $V'$ .

Note that since  $x_0 \notin V'$ , this square submatrix  $L'$  is contained in  $L(D)^{x_0, x_0}$ , and one has a block decomposition:

$$L(D) = \begin{array}{c} \left. \begin{array}{c} \underbrace{\hspace{10em}}_{v'} \\ \underbrace{\hspace{10em}}_{v-v'} \end{array} \right\} \begin{array}{c} \left[ \begin{array}{cc|c} L' & 0 & \\ \hline * & * & \end{array} \right] \end{array}$$

← Why are these entries all 0?

### EXAMPLE



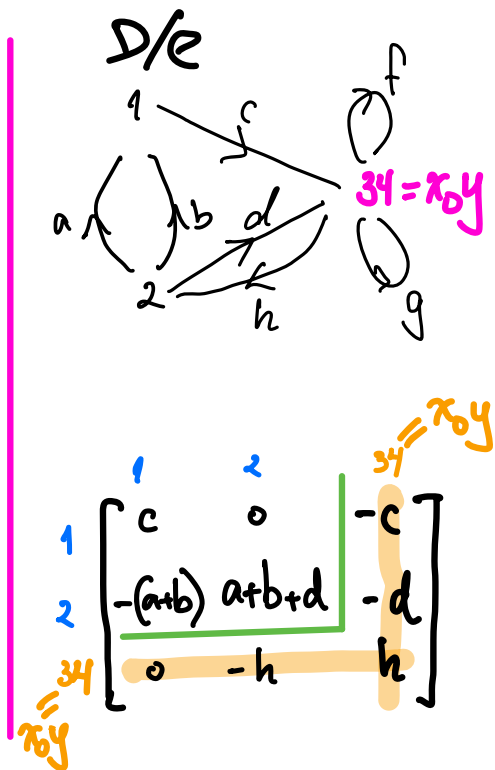
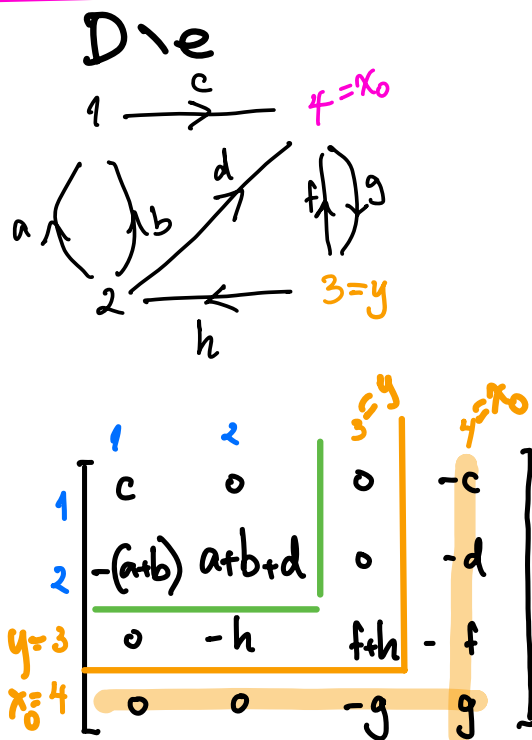
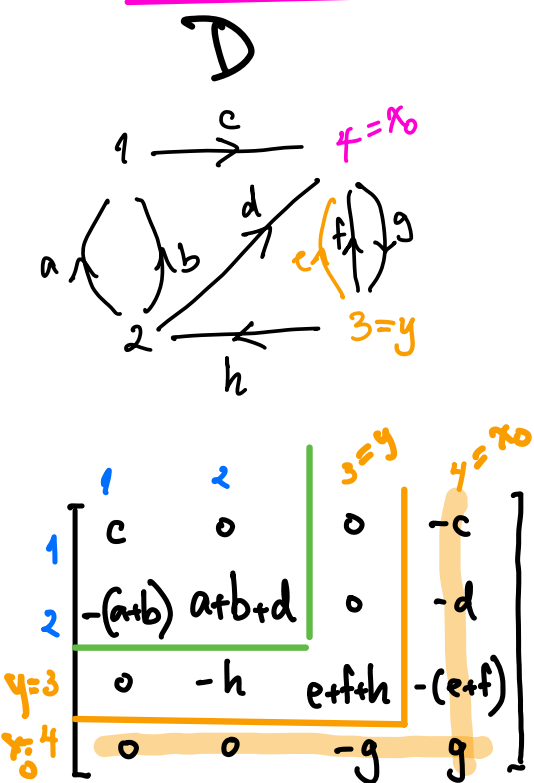
$$L(D) = \begin{array}{c} \left. \begin{array}{c} \underbrace{\hspace{10em}}_{v'} \\ \underbrace{\hspace{10em}}_{v-v'} \end{array} \right\} \begin{array}{c} \left[ \begin{array}{ccc|cc} 1 & 2 & 3 & 4 & 5 \\ \hline b+g & -b & -g & 0 & 0 \\ -a & a & 0 & 0 & 0 \\ 0 & -c & c & 0 & 0 \\ \hline 0 & -d & 0 & d+e & -e \\ 0 & 0 & 0 & -f & f \end{array} \right] \end{array}$$

Then  $\det(L') = 0$  because every row of  $L'$  sums to 0, so the all 1's vector is in its nullspace.

Therefore  $\det(L(D)^{x_0, x_0}) = \underbrace{\det(L')}_{=0} \cdot \det(L(D)^{v-v'-x_0, v-v'-x_0}) = 0$ .

This checks (a), (b), (c) for  $\det(L(D)^{\kappa_0, \kappa_0})$ , completing the base cases for the induction on  $|A|$ . In the inductive step, one may assume  $\exists$  some arc  $e = (y, \kappa_0)$ .

Consider  $L(D)^{\kappa_0, \kappa_0}$ ,  $L(D \setminus e)^{\kappa_0, \kappa_0}$ ,  $L(D/e)^{\kappa_0 y, \kappa_0 y}$ :



$$\det L(D)^{\kappa_0, \kappa_0} = \det L(D \setminus e)^{\kappa_0, \kappa_0} + e \cdot \det L(D/e)^{\kappa_0 y, \kappa_0 y}$$

↑  
by expansion  
along the  $y$  column

$$= t(D \setminus e, \kappa_0; e) + e \cdot t(D/e, \kappa_0; e)$$

by induction on  $|A|$

$$= t(D, \kappa_0; e)$$

by (d)



# Consequences of the Matrix-Tree Theorem

① Computing the integers  $t(G)$  or  $t(D)$  can be done in  $\leq c \cdot N^3$  steps where  $N=|V|$  via **linear algebra**. One can use Gaussian elimination/row-reduction to compute  $\det L(G)^{x_0, x_0}$  or  $\det L(D)^{x_0, x_0}$  recalling that ...

- adding multiples of a row of  $A$  to another leaves  $\det A$  unchanged
- swapping two rows of  $A$  negates  $\det A$
- scaling a row of  $A$  by  $c$  also scales  $\det A$  by  $c$ .

EXAMPLE:

$$\det \begin{bmatrix} 2 & 3 & 4 \\ 4 & 6 & 2 \\ 3 & 4 & 5 \end{bmatrix} = \det \begin{bmatrix} 2 & 3 & 4 \\ 0 & 0 & -6 \\ 3 & 4 & 5 \end{bmatrix} = - \det \begin{bmatrix} 2 & 3 & 4 \\ 3 & 4 & 5 \\ 0 & 0 & -6 \end{bmatrix}$$

*subtract  $-2(\text{row } 1)$  from row 2*      *permute 2 rows*

$$= -2 \det \begin{bmatrix} 1 & 3/2 & 2 \\ 3 & 4 & 5 \\ 0 & 0 & -6 \end{bmatrix} = -2 \det \begin{bmatrix} 1 & 3/2 & 2 \\ 0 & -1/2 & -1 \\ 0 & 0 & -6 \end{bmatrix} = -2(1)(-1/2)(-6) = -6$$

*scale row 1*      *subtract  $3(\text{row } 1)$  from row 2*

② Occasionally one can evaluate  $t(G)$  theoretically via **eigenvalues of  $L(G)^{\chi_0, \chi_0}$**

since if  $L(G)^{\chi_0, \chi_0} = P^{-1} \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix} P$


then  $\det L(G)^{\chi_0, \chi_0} = \det P^{-1} \cdot \det \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix} \cdot \det P$   
 $= \lambda_1 \lambda_2 \dots \lambda_n$

---

EXAMPLE:

**THEOREM:**  $t(K_n) = n^{n-2}$   
 (Cayley, 1889; Borchardt, 1860)

e.g.  $t(K_3) = t\left(\begin{smallmatrix} 1 & & \\ & 1 & \\ & & -3 \end{smallmatrix}\right) = 3 = 3^{3-2}$



$t(K_4) = t\left(\begin{smallmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & -3 \end{smallmatrix}\right) = 16 = 4^{4-2}$



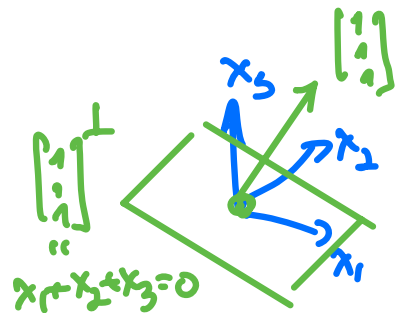
proof 1: Find the eigenvalues of  $L(K_n^{x_0, x_0})$

$$L(K_n) = \begin{matrix} & \begin{matrix} 1 & 2 & \dots & n-1 & n \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ \vdots \\ n-1 \\ n \end{matrix} & \begin{bmatrix} n-1 & -1 & & & -1 \\ -1 & n-1 & & & -1 \\ & & \ddots & & \vdots \\ -1 & -1 & \dots & n-1 & -1 \\ -1 & -1 & \dots & -1 & n-1 \end{bmatrix} \end{matrix}$$

$$L(K_n^{x_0, x_0}) = \begin{matrix} & \begin{matrix} 1 & 2 & \dots & n-1 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ \vdots \\ n-1 \end{matrix} & \begin{bmatrix} n-1 & & & \\ & n-1 & & \\ & & \ddots & \\ -1 & & & n-1 \end{bmatrix} \end{matrix} = n \cdot \mathbb{I}_{n-1} - \underbrace{\begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}}_{\mathbb{J}_{n-1} := \text{all ones matrix } (n-1) \times (n-1)}$$

Note  $\mathbb{J}_{n-1}$  has an eigenvector  $\begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$  with eigenvalue  $n-1$ :

$$n-1 \cdot \left\{ \underbrace{\begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}}_{n-1} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} n-1 \\ n-1 \\ \vdots \\ n-1 \end{bmatrix} = (n-1) \cdot \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \right.$$



Also the  $(n-2)$ -dimensional perpendicular space

$$\begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}^\perp = \left\{ \mathbf{x} \in \mathbb{R}^{n-1} : x_1 + x_2 + \dots + x_{n-1} = 0 \right\}$$

lies in the nullspace (= 0-eigenspace) of  $\mathbb{J}_{n-1}$ :

$$\begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_{n-1} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = 0 \cdot \begin{bmatrix} x_1 \\ \vdots \\ x_{n-1} \end{bmatrix}$$

Hence  $J_{n-1}$  has eigenvalues  $(\lambda_1, \lambda_2, \dots, \lambda_{n-2}, \lambda_{n-1})$   
 $= (0, 0, \dots, 0, n-1)$

and therefore

$L(K_n)^{x_0, x_0} = nI_n - J_{n-1}$  has eigenvalues  
 $(n-\lambda_1, n-\lambda_2, \dots, n-\lambda_{n-2}, n-\lambda_{n-1})$   
 $= (n, n, \dots, n, 1)$

and  $\det L(K_n)^{x_0, x_0} = \underbrace{n \cdot n \dots n}_{n-2 \text{ times}} \cdot 1 = n^{n-2} \quad \square$

proof 2: There is a beautiful bijection called **Prüfer coding** (1918)

$\{\text{spanning trees } T \text{ in } K_n\} \xrightarrow{c} \{1, 2, \dots, n\}^{n-2}$   
 $= \{(c_1, c_2, \dots, c_{n-2}) : c_i \in \{1, 2, \dots, n\}\}$   
 a set of size  $n^{n-2}$

$T \mapsto c(T) = (c_1, c_2, \dots, c_{n-2})$

defined by letting  $c_1 =$  unique neighbor of smallest labeled leaf  $l_1$  of  $T$

$c_2 =$  neighbor of smallest leaf of  $T - \{l_1\}$

$c_3 =$  neighbor of smallest leaf of  $T - \{l_1, l_2\}$

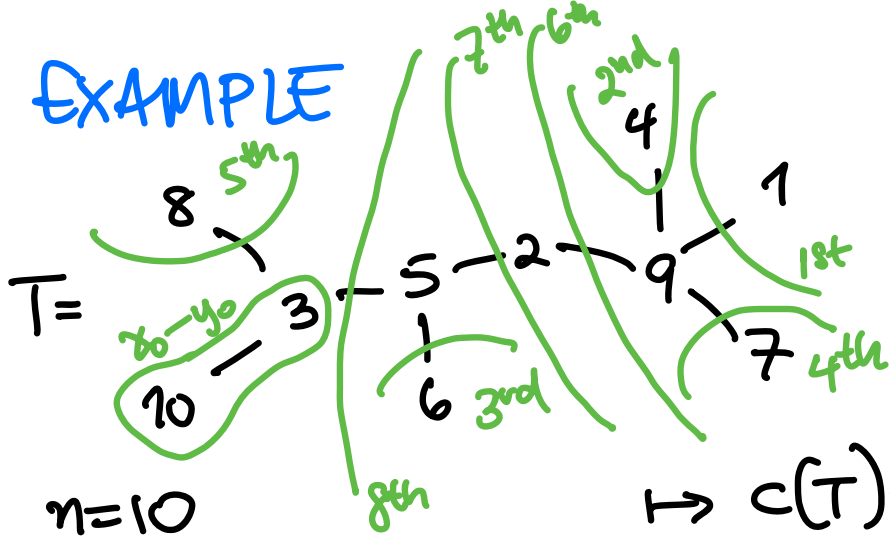
$\vdots$

$c_{n-2} = \dots$

and stopping at the **edge**  $T - \{l_1, l_2, \dots, l_{n-2}\} = x_0 - y_0$

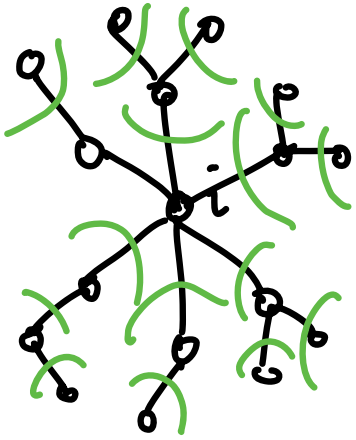


# EXAMPLE



$$\begin{aligned}
 \mapsto c(T) &= (c_1, c_2, c_3, c_4, c_5, c_6, c_7, c_8) \\
 &= (9, 9, 5, 9, 3, 2, 5, 3)
 \end{aligned}$$

The inverse map  $\bar{c}^{-1}$  takes advantage of this



FACT:  $d_T(i) =$

$$1 + \#\{\text{occurrences of } i \text{ in } c(T)\}$$

In particular,  $i$  is a leaf vertex of  $T$

$\iff i$  does not appear in  $c(T)$

One uses this to recover the edges of  $T$  by creating  $l_i - c_i$  edges and crossing off leaves. Vertices disappearing from the code get added to the leaf list.

# EXAMPLE

$c(T - \{l_1, l_2, \dots, l_i\})$

leaves of  $T - \{l_1, l_2, \dots, l_i\}$

(9, 9, 5, 9, 3, 2, 5, 3)

1, 4, 6, 7, 8, 10

(9, 5, 9, 3, 2, 5, 3)

4, 6, 7, 8, 10

(5, 9, 3, 2, 5, 3)

6, 7, 8, 10

(9, 3, 2, 5, 3)

7, 8, 10

(3, 2, 5, 3)

8, 9, 10

(2, 5, 3)

9, 10

(5, 3)

2, 10

(3)

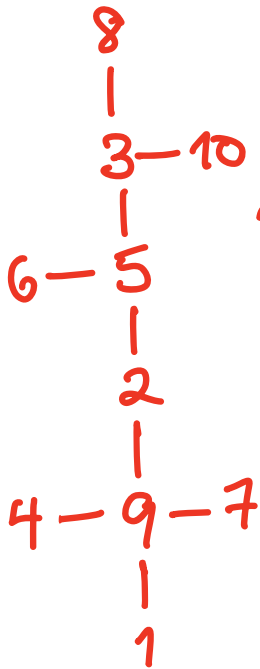
5, 10

()

3, 10

$x_0 - y_0$

← assemble the edges to form T



## ACTIVE LEARNING<sub>1</sub>:

Someone picks a random  $c = (c_1, c_2, \dots, c_7) \in \{1, 2, \dots, 9\}^7$   
 and we'll all compute  $T = c^{-1}(c)$

proof (that Prüfer coding is a bijection):

One can see that  $\bar{c}^{-1}(c(T)) = T$ , but how do we know that for any  $\underline{c} \in \{1, 2, \dots, n\}^{n-2}$ , the (multi-)graph  $G$  produced by  $\bar{c}^{-1}$  from  $\underline{c}$  really is a tree. This follows by working backwards, adding in edge  $\{x_0, y_0\}$ , then  $\{c_{n-2}, l_{n-2}\}$ , then  $\{c_{n-3}, l_{n-3}\}, \dots, \{c_1, l_1\}$ . At each stage, can check  $l_i$  gets connected to the tree containing  $\{x_0, y_0\}$  being built up, and was isolated before that. So every vertex has a path to  $\{x_0, y_0\}$  and no cycles every get created.  $\square$

**COROLLARY** (to Prüfer coding)

The number of spanning trees  $T$  in  $K_n$  with

$$d(T) = (d_1, d_2, \dots, d_n) \quad (\text{where } d_1 + d_2 + \dots + d_n = \underbrace{2(n-1)}_{\text{why?}})$$

is the **multinomial coefficient**

$$\binom{n-2}{d_1-1, d_2-1, \dots, d_n-1} = \frac{(n-2)!}{(d_1-1)! (d_2-1)! \dots (d_n-1)!} = \text{coefficient of } x_1^{d_1-1} x_2^{d_2-1} \dots x_n^{d_n-1} \text{ in } (x_1 + x_2 + \dots + x_n)^{n-2}$$

Equivalently, 
$$\sum_{\text{spanning trees } T \text{ in } K_n} x_1^{d_T(1)} x_2^{d_T(2)} \dots x_n^{d_T(n)} = x_1 x_2 \dots x_n (x_1 + x_2 + \dots + x_n)^{n-2}$$

# EXAMPLES

①

$$\sum_{\text{spanning trees } T \text{ in } K_3} d_T(1) d_T(2) d_T(3) = x_1 x_2^2 x_3 + x_1 x_2 x_3^2 + x_1^2 x_2 x_3$$

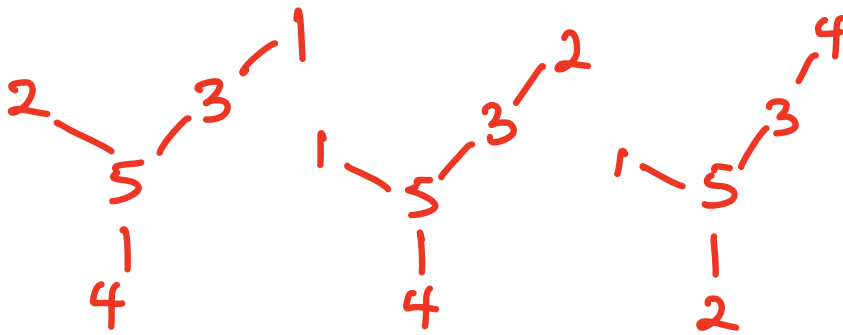
$$= x_1 x_2 x_3 (x_1 + x_2 + x_3)^{3-2}$$



② How many spanning trees  $T$  in  $K_5$  have

$$d(T) = (1, 1, 2, 1, 3) ? \quad \binom{3}{1, 1, 2, 1, 3-1} =$$

$$\binom{3}{0, 0, 1, 1, 2} =$$



$$\frac{3!}{0! 0! 1! 1! 2!} =$$

$$\frac{3 \cdot 2 \cdot 1}{1 \cdot 1 \cdot 1 \cdot 1 \cdot 2} = 3$$

→ their Prüfer codes  $(c_1, c_2, c_3)$

will be rearrangements of  
 zero 1's  
 zero 2's  
 one 3  
 zero 4's  
 two 5's

$$\text{i.e. } \{ (3, 5, 5), (5, 3, 5), (5, 5, 3) \}$$