



Tensors

Assume  $A$  is  $k \times l$  &  $B$  is  $m \times n$ . The **Kronecker product**  $A \otimes B$  is the  $km \times ln$  block matrix with  $ij$ -block equal to  $A_{ij}B$ .

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \otimes \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} a & 2a \\ b & 2b \\ c & 2c \\ 3a & 4a \\ 3b & 4b \\ 3c & 4c \end{pmatrix}$$

## Properties

1)  $A \otimes B$  is linear in each term

2) If the required products exist

$$(A \otimes B)(C \otimes D) = A(C) \otimes B(D)$$

$$(A \otimes B) \circ (C \otimes D) = (A \circ C) \otimes (B \circ D)$$

3) If  $Ax = \lambda x$ ,  $By = \mu y$ , then  $(A \otimes B)(x \otimes y) = \lambda \mu (x \otimes y)$

4) If  $P: u \otimes v \rightarrow v \otimes u$  ( $\forall u, v$ ) then

$P(M \otimes M^T)$  is symmetric.

A map  $\text{Mat}_{m \times n}(\mathbb{R}) \rightarrow \mathbb{R}^{mn}$

If  $A \in \text{Mat}_{m \times n}(\mathbb{R})$ , then

$$\text{vec}(A) = \begin{bmatrix} Ae_1 \\ \vdots \\ Ae_n \end{bmatrix}$$

This is clearly linear.

**Lemma**  $\text{vec}(BMA^T) = (A \otimes B) \text{vec}(M)$

## Tensor product of modules

$M, N$  modules over ring  $R$ .

Construct module  $M \otimes N$  as quotient of finitely supported functions in  $R^{M \times N}$  over ideal generated by relations:

$$(x, y+z) - (x, y) - (x, z), (w+x, y) - (w, y) - (x, y)$$

$$(ax, y) - a(x, y), (x, by) - b(x, y)$$

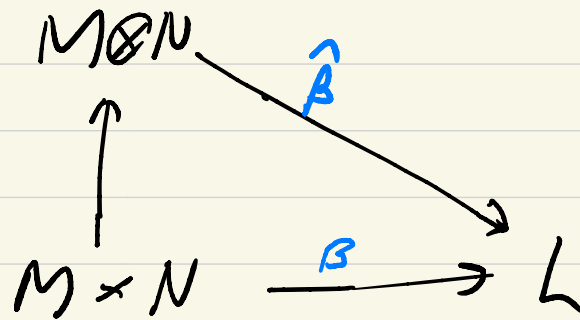
Image of  $(x, y)$  in quotient is  $x \otimes y$

The Kronecker product is the  
tensor product on vector spaces

## Bilinear maps

The tensor product turns questions about bilinear maps on  $M \times N$  to questions about linear maps.

Moral: complicate the objects, keep the maps simple:





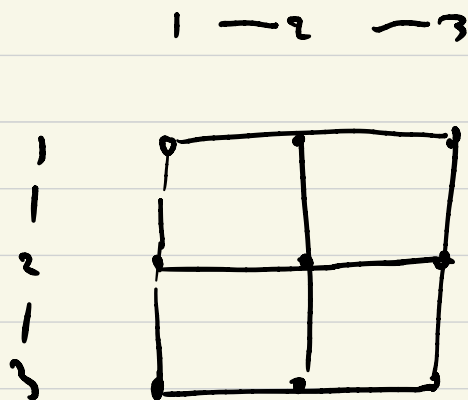
Graph products

Cartesian product  $X \square Y$

$$A(X \square Y) = A(X) \otimes I + I \otimes A(Y)$$

$$\text{dist}((u,v), (x,y))$$

$$= \text{dist}(u,x) + \text{dist}(v,y)$$

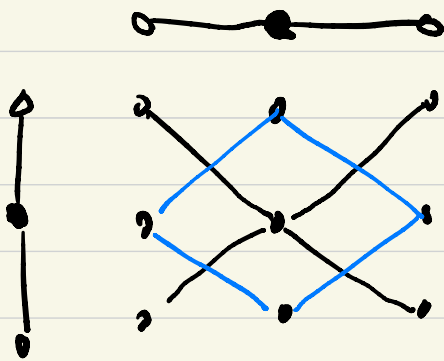


eigenvalues  $\lambda_r + \mu_s$   $\forall r, s$

# Direct product $X \times Y$

$$A(X \times Y) = A(X) \otimes A(Y)$$

(a)

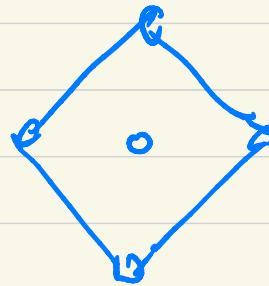
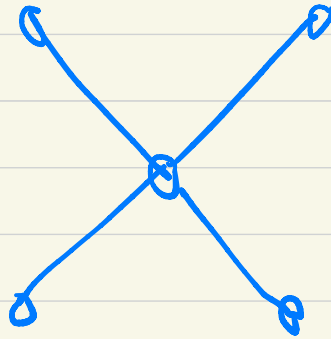


eigenvalues  $\lambda_r \mu_s \forall r, s$

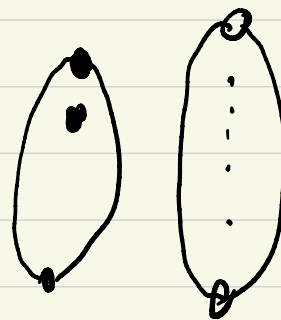
$\cdot$   $X, Y$  connected  
 $X \times Y$  connected  
unless  $X$  &  $Y$  both  
bipartite

$$(b) \quad K_2 \otimes 2K_3 \cong 2C_6 \cong K_2 \otimes C_6$$

# Cospectral graphs I



What is the spectrum of  $K_{m,n}$ ?



# Components of tensor product

$$\begin{pmatrix} 0 & B^T \\ B & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & C^T \\ C & 0 \end{pmatrix} = \begin{bmatrix} 0 & 0 & 0 & B^T C^T \\ 0 & 0 & B^T C & 0 \\ \hline 0 & B C^T & 0 & 0 \\ B C & 0 & 0 & 0 \end{bmatrix}$$

$$\sim \begin{bmatrix} 0 & B^T C^T & 0 & 0 \\ B C^T & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & B^T C^T \\ 0 & 0 & B C & 0 \end{bmatrix}$$

$$A(P_3) = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} = \begin{pmatrix} C & B^T \\ B & A \end{pmatrix}$$

$$\begin{pmatrix} 0 & B^T \otimes B^T \\ B \otimes B & 0 \end{pmatrix} \quad \begin{pmatrix} 0 & B^T \otimes B \\ B \otimes B^T & 0 \end{pmatrix}$$

C

D

$$\begin{bmatrix} 0 & 1 & 1 & 1 \\ \vdots & & & \\ \vdots & & & \\ \vdots & & & \\ \vdots & & & \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \end{bmatrix}$$

$$C^2 = \begin{pmatrix} B^T \otimes B^T & 0 \\ 0 & B \otimes B \end{pmatrix}, \quad D^2 = \begin{bmatrix} B^T \otimes B^T & 0 \\ 0 & B \otimes B \end{bmatrix}$$

**Theorem** If  $A$  is  $m \times n$  &  $B$  is  $n \times m$ ,

$$\det(I - AB) = \det(I - BA)$$

Proof.

$$\begin{pmatrix} I & 0 \\ -B & I \end{pmatrix} \begin{pmatrix} I & A \\ B & I \end{pmatrix} = \begin{pmatrix} I & A \\ 0 & I - AB \end{pmatrix}$$

$$\begin{pmatrix} I & A \\ B & I \end{pmatrix} \begin{pmatrix} I & 0 \\ -B & I \end{pmatrix} = \begin{pmatrix} I - AB & A \\ 0 & I \end{pmatrix}$$

Now take determinants.

□



**Corollary**  $t^{m-n} \det(tI - AB) = \det(tI - BA)$

If  $A$  is  $m \times n$  &  $B$  is  $n \times m$ , then

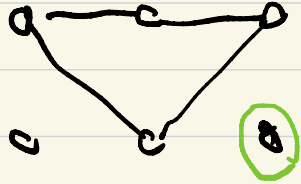
**Corollary** If  $A$  is  $m \times n$  and  $B$  is  $n \times m$ , then

$AB, BA$

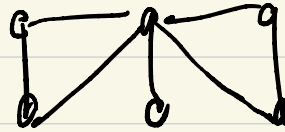
have the same non-zero eigenvalues, with the same multiplicities.

**Lemma** If  $X$  and  $Y$  are bipartite graphs, the two components<sup>?</sup> of  $X \times Y$  have the same non-zero eigenvalues, with the same multiplicities

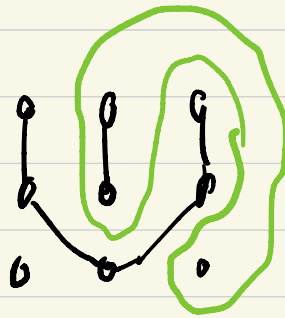
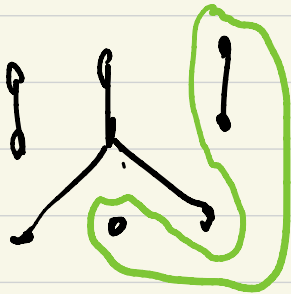
Partitioned tensor product



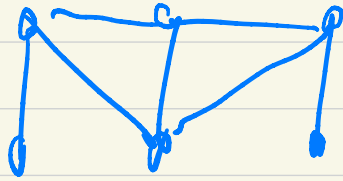
smallest pair of  
cospectral graphs



smallest pair of  
connected cospectral  
graphs



smallest pair of  
cospectral forests



$$\begin{pmatrix} A_1 & H \\ H^T & A_2 \end{pmatrix}$$

$$\begin{pmatrix} A_2 & H^T \\ H & A_1 \end{pmatrix}$$

$$\begin{pmatrix} I & B \\ B^T & I \end{pmatrix} = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} I & B \\ B^T & I \end{pmatrix} \otimes \begin{pmatrix} A_1 & H \\ H^T & A_2 \end{pmatrix} = \begin{pmatrix} I \otimes A_1 & B \otimes H \\ B^T \otimes H^T & I \otimes A_2 \end{pmatrix}$$

partitioned  
tensor  
product

$$\begin{pmatrix} I & B \\ B^T & I \end{pmatrix} \otimes \begin{pmatrix} A_2 & H^T \\ H & A_1 \end{pmatrix} = \begin{pmatrix} I \otimes A_2 & B \otimes H^T \\ B^T \otimes H & I \otimes A_1 \end{pmatrix}$$

Choose  $Q$  &  $R$  so  $Q^T A = I$ ,  $R^T R = I$  and

$$Q^T B R = \Sigma - \text{diagonal}$$

How?

$$\begin{pmatrix} Q^T \otimes I & 0 \\ 0 & R^T \otimes I \end{pmatrix} \begin{pmatrix} I \otimes A_1 & B \otimes H \\ B^T \otimes H^T & I \otimes A_2 \end{pmatrix} \begin{pmatrix} Q \otimes I & 0 \\ 0 & R \otimes I \end{pmatrix}$$

$$= \begin{pmatrix} I \otimes A_1 & \Sigma \otimes H \\ \Sigma \otimes H^T & I \otimes A_2 \end{pmatrix}$$

e.g.  $B$  is  $2 \times 3$ :

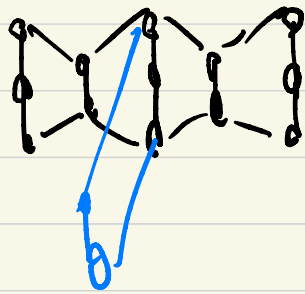
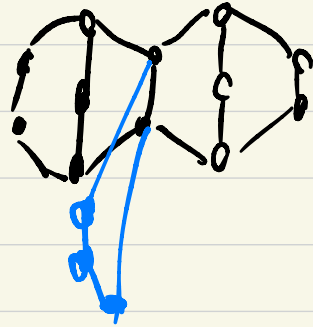
$$\begin{pmatrix} I \otimes A_1 & B \otimes H \\ B^T \otimes H^T & I \otimes A_2 \end{pmatrix} = \left[ \begin{array}{cc|ccc} A_1 & 0 & \sigma_1 H & 0 & 0 \\ 0 & A_1 & 0 & \sigma_2 H & 0 \\ \hline \sigma_1 H^T & 0 & A_2 & 0 & 0 \\ 0 & \sigma_2 H^T & 0 & A_2 & 0 \\ 0 & 0 & 0 & 0 & A_2 \end{array} \right] \sim \left[ \begin{array}{cc|cc|c} A_1 & \sigma_1 H & 0 & 0 & 0 \\ \sigma_1 H^T & A_2 & 0 & 0 & 0 \\ \hline 0 & 0 & A_1 & \sigma_2 H & 0 \\ 0 & 0 & \sigma_2 H^T & A_2 & 0 \\ \hline 0 & 0 & 0 & 0 & A_2 \end{array} \right]$$

$$\begin{pmatrix} I \otimes A_2 & B \otimes H^T \\ B^T \otimes I & I \otimes A_1 \end{pmatrix} = \begin{pmatrix} A_2 & 0 & \sigma_1 H^T & 0 & 0 \\ 0 & A_2 & 0 & \sigma_2 H^T & 0 \\ \sigma_1 H & 0 & A_1 & 0 & 0 \\ 0 & \sigma_2 H & 0 & A_1 & 0 \\ 0 & 0 & 0 & 0 & A_1 \end{pmatrix} \sim \begin{pmatrix} A_2 & \sigma_1 H^T & 0 & 0 & 0 \\ \sigma_1 H & A_1 & 0 & 0 & 0 \\ 0 & 0 & A_1 & \sigma_2 H^T & 0 \\ 0 & 0 & \sigma_2 H & A_2 & 0 \\ 0 & 0 & 0 & 0 & A_1 \end{pmatrix}$$

(proof due to  
Knuth)

$$\begin{bmatrix} A_1 & H \\ H^T & A_2 \end{bmatrix} \sim \begin{bmatrix} A_2 & H^T \\ H & A_1 \end{bmatrix}$$





$$G \rightarrow H \leftarrow G \rightarrow H \leftarrow G$$

$$\quad \quad \quad \downarrow$$

$$\quad \quad \quad H$$

$$H \leftarrow G \rightarrow H \leftarrow G \rightarrow H$$

$$\quad \quad \quad \uparrow$$

$$\quad \quad \quad G$$

Singular values

**Problem** Find the best rank-1 approximation  
to  $B$ .

Assume  $\|y\| = \|z\| = 1$ . Then

$$\langle B - \lambda yz^T, B - \lambda yz^T \rangle = \langle B, B \rangle - 2y^T B z + \lambda^2$$

Minimum occurs when  $\lambda = y^T B z$ ,

value is  $\langle B, B \rangle - (y^T B z)^2$ .

We need to maximize  $(y^T B z)^2$

Suppose  $h \in y^\perp$ ,  $\|h\|$  small

$$(y^T B z + h^T B z)^2 \approx (y^T B z)^2 + (y^T B z)(h^T B z)$$

Hence if  $y$  maximizes,  $h^T B z = 0$  and  $h \in (B z)^\perp$ .

So  $y^\perp \in (B z)^\perp$  &  $B z \in \langle y \rangle$

Therefore if  $yz^T$  is optimal

$$Bz = \lambda y, \quad B^T y = \mu z \quad \Rightarrow \begin{aligned} y &\in \text{col}(B) \\ z &\in \text{col}(B^T) \end{aligned}$$

and

$$\lambda \mu z = \lambda B^T y = B^T B z$$

$$\mu \lambda y = \mu B^T z = B B^T y$$

Also

$$\lambda = y^T B z; \quad \mu = z^T B y \quad \Rightarrow \lambda = \mu$$

Traditionally we set  $\sigma$  equal to  $\lambda, \mu$

**Theorem** Assume  $B$  is  $m \times n$  &  $\text{rk}(B) = k$ .

Let  $\sigma_1 \geq \dots \geq \sigma_k$  be the square roots of the eigenvalues of  $BB^T$ . Then there is an orthonormal basis  $y_1, \dots, y_k$  of  $\text{col}(B)$  and an orthonormal basis  $z_1, \dots, z_r$  of  $\text{col}(B^T)$  such

that

$$B = \sum_{r=1}^k \sigma_r y_r z_r^T$$

$$Q := [y_1, \dots, y_k], \quad R := [\beta_1, \dots, \beta_k], \quad \Sigma := \text{diag}(\sigma_i)$$

$$B = Q \Sigma R^T$$

$m \times n$       $m \times k$       $k \times n$

$$Q^T Q = I, \quad R^T R = I$$

singular value decomposition

$$(\text{See } Q^T B R = \Sigma.)$$



