

# Math 221 Lec 11

## 2.4: Elementary matrices, 2.5: Transpose

Asa Royal (ajr74)

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## 1 Transpose

**Remark.** The transpose of a matrix  $A$  is the matrix  $A^T$  where  $(A^T)_{ij} = A_{ji}$ . Entries are reflected over the diagonal

**Remark.** If  $x, y \in \mathbb{R}^n$ , we can think of their dot dot product  $\mathbf{x} \cdot \mathbf{y}$  as

$$\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \mathbf{y}$$

**Remark.** The remark above reveals that vectors themselves are linear transformations.  $\mathbf{x} \in \mathbb{R}^n$  is  $n$ -dimensional, but  $\mathbf{x} \cdot \mathbf{y}$  is 1-dimensional, so  $\mathbf{x}^T$  is the  $1 \times n$  matrix that encodes a linear transformation that takes an  $n$ -dimensional vector to 1 dimension.  $\mathbf{x}_1$  tells us what the linear transformation maps  $\hat{i}$  to, and so on... Add all that hat vectors and you get a scalar: the output of the linear transformation.

**Definition 1 (dual).** The **dual** of a vector is the linear transformation it encodes. The **dual** of a linear transformation from  $n$  dimensions to 1 dimension is a certain vector.

**Remark.** Note then that  $\mathbf{a}^T \mathbf{a} = \mathbf{a} \cdot \mathbf{a} = \|\mathbf{a}\|^2$

**Remark.**  $\mathbf{x} \cdot \mathbf{y}$  produces a scalar, so  $\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \mathbf{y}$  or  $\mathbf{y}^T \mathbf{x}$  (both of which are  $1 \times 1$  matrices), but  $\mathbf{x} \cdot \mathbf{y} \neq \mathbf{y} \mathbf{x}^T$  or  $\mathbf{x} \mathbf{y}^T$ , both of which would be  $n \times n$  matrices.

**Remark.** The linear transformation represented by the  $n \times n$  matrix  $\mathbf{a} \mathbf{a}^T$  is  $\text{proj}_{\mathbf{a}} \mathbf{x}$ . By expressing the projection formula in terms of  $\mathbf{a}$  and  $\mathbf{a}^T$ , we can clearly show that it is a function in terms of  $\mathbf{a}$ .

*Proof.*

$$\begin{aligned} \text{proj}_{\mathbf{a}} \mathbf{x} &= \frac{\mathbf{x} \cdot \mathbf{a}}{\|\mathbf{a}\|^2} \mathbf{a} \\ &= \frac{\mathbf{a}^T \mathbf{x}}{\|\mathbf{a}\|^2} \mathbf{a} && \text{(express dot product as matrix mult, per second remark above)} \\ &= \mathbf{a} \frac{\mathbf{a}^T \mathbf{x}}{\|\mathbf{a}\|^2} && \text{(can move since one of the terms above is a scalar)} \\ &= \frac{\mathbf{a} \mathbf{a}^T \mathbf{x}}{\mathbf{a} \cdot \mathbf{a}} \\ &= \left( \frac{\mathbf{a} \mathbf{a}^T}{\mathbf{a}^T \mathbf{a}} \right) \mathbf{x} && \text{(again, express dot product as matrix mult)} \end{aligned}$$

□

**Example** (projection expressed with transposes).

$$\begin{aligned} \text{proj} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} &= \frac{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}}{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}} && \text{denom is } 1 \times n \text{ times } n \times 1, \text{ aka dot prod} \\ &= \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \end{aligned}$$

**Definition 2** (orthogonal matrix). An  $n \times n$  matrix  $A$  is orthogonal if  $A^T A = I_n$ , which is true iff  $\mathbf{a}_i \cdot \mathbf{a}_j = 0$  for  $i \neq j$  and 1 for  $i = j$ .

## 2 Left multiplication (row vector \* matrix)

**Remark.** Let  $A$  be an  $n \times m$  matrix. Let  $\mathbf{x}$  be a vector of length  $n$ . The operation in which a row vector is multiplied by a matrix can be expressed as  $\mathbf{x}^T A$ .

**Proposition 3.**  $A\mathbf{x} \cdot \mathbf{y} = \mathbf{x} \cdot A^T \cdot \mathbf{y}$  (To move the matrix across a dot product operator, we must transpose it)

*Proof.*

$$\begin{aligned} (A\mathbf{x}) \cdot \mathbf{y} &= (A\mathbf{x})^T \mathbf{y} && \text{dot product as matrix mult} \\ &= \mathbf{x}^T A^T \mathbf{y} \\ &= \mathbf{x}^T (A^T \mathbf{y}) \\ &= \mathbf{x} \cdot A^T \mathbf{y} \end{aligned}$$

□

## 3 Elementary matrices

**Proposition 4.** Every invertible matrix can be expressed as a product of elementary matrices.

*Proof.* Since  $A$  is invertible, we can row reduce it to the identity matrix. We do this by multiplying  $A$  on the left by a series of elementary matrices  $E = (E_k)(\dots)(E_2)(E_1)$  such that  $EA = I$ . Elementary matrices are invertible, so we can multiply both sides of that equation by  $E^{-1}$ . Then  $A = E^{-1} = (E_1^{-1})(E_2^{-1})(\dots)(E_k^{-1})$ . □

**Remark.** When we apply elementary operations to the rows of a matrix  $A$ , we multiply  $A$  on the left by an elementary matrix  $E$ , such that we get a transformed version of the rows of  $A$ .

$$EA = \begin{bmatrix} - & E_1 A & - \\ - & E_2 A & - \\ & \vdots & \\ - & E_m A & - \end{bmatrix}$$

**Remark.** We construct a row swap elementary matrix  $E$  by taking  $I_m$  and interchanging rows  $i$  and  $j$  to swap  $A_i$  and  $A_j$ . A row of  $E$ ,  $E_k$ , looks like

$$\begin{cases} e_k^T, & \text{if } k \neq i \text{ or } j \\ e_j^T, & \text{if } k = i \\ e_i^T, & \text{if } k = j \end{cases}$$

where  $e_i$  is the  $i$ -th basis vector and  $e_i^T$  is the  $i$ -th basis row vector.

Thus,

$$E_k A = \begin{cases} A_k, & \text{if } k \neq i \text{ or } j \\ A_j, & \text{if } k = i \\ A_i, & \text{if } k = j \end{cases}$$