Linear Algebra Refresher

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**Vector** - We note  $x \in \mathbb{R}^n$  a vector with *n* entries, where  $x_i \in \mathbb{R}$  is the *i*<sup>th</sup> entry:

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**Matrix** - We note  $A \in \mathbb{R}^{m \times n}$  a matrix with m rows and n columns, where  $A_{i,j} \in \mathbb{R}$  is the entry located in the  $i^{\text{th}}$  row and  $j^{\text{th}}$  column

$$
A = \begin{pmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{pmatrix} \in \mathbb{R}^{m \times n}
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**Identity matrix** - The identity matrix  $I \in \mathbb{R}^{n \times n}$  is a square matrix with ones in its diagonal and zero everywhere else

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I = \left( \begin{array}{cccc} 1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & 1 \end{array} \right)
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**Diagonal matrix** - A diagonal matrix  $D \in \mathbb{R}^{n \times n}$  is a square matrix with nonzero values in its diagonal and zero everywhere else:

$$
D = \left( \begin{array}{cccc} d_1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & d_n \end{array} \right)
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outer product: for  $x \in \mathbb{R}^m$ ,  $y \in \mathbb{R}^n$ , we have:

$$
xy^T = \left(\begin{array}{ccc} x_1y_1 & \cdots & x_1y_n \\ \vdots & & \vdots \\ x_my_1 & \cdots & x_my_n \end{array}\right) \in \mathbb{R}^{m \times n}
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Matrix-vector multiplication - The product of matrix  $A \in \mathbb{R}^{m \times n}$  and vector  $x \in \mathbb{R}^n$  is a vector y of size  $\mathbb{R}^m$ , such that for all  $i \in \{1, \ldots, m\}$ :

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**Matrix-matrix multiplication** - The product of matrix  $A \in \mathbb{R}^{m \times n}$  and matrix  $B \in \mathbb{R}^{n \times p}$  is a matrix C of size  $\mathbb{R}^{m \times p}$ , such that for all  $i \in \{1, \ldots, m\}$  and all  $j \in \{1, \ldots, p\}$ :

$$
C_{ij}=\sum_{k=1}^n A_{ik}B_{kj}.
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**Transpose** - The transpose of a matrix is an operator which flips a matrix over its diagonal. Formally, the transpose of a matrix  $A \in \mathbb{R}^{m \times n}$ , noted  $A^{\mathcal{T}}$ , is defined, for all  $i \in \{1, \ldots, m\}$  and all  $j \in \{1, \ldots, n\}$ , as

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**Inverse** -  $A \in \mathbb{R}^{n \times n}$  is said to be invertible if there exists  $B \in \mathbb{R}^{n \times n}$  such that  $AB = BA = I_n$ . If this is the case, then the matrix B is uniquely determined by A, and is called the inverse of A, denoted  $A^{-1}$ .

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**Trace** - The trace of a square matrix  $A \in \mathbb{R}^{n \times n}$ , noted tr(A), is the sum of its diagonal entries

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**Determinant** - The determinant of a matrix A, commonly denoted  $det(A)$ , characterizes some properties of the matrix. In particular, the determinant is nonzero if and only if the matrix is invertible.

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For matrices  $A, B$ , we have

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(AB)^{T} = B^{T}A^{T}.
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For square matrices  $A, B$ , we have

 $det(A^T) = det(A)$  and  $det(AB) = det(A) det(B)$ .