2.3 Operations on Linear Transformations and Matrices

Definitions: If
$$T_1 : \mathbb{R}^n \to \mathbb{R}^m$$
 and $T_2 : \mathbb{R}^n \to \mathbb{R}^m$ are linear
transformations, and $k \in \mathbb{R}$, then we can define:
 $Sum of T_1 \& T_2$
 $T_1 + T_2 : \mathbb{R}^n \to \mathbb{R}^m$,
 $T_1 - T_2 : \mathbb{R}^n \to \mathbb{R}^m$, and
 $Scale c hadred of kT_1 : \mathbb{R}^n \to \mathbb{R}^m$,
as linear transformations, with actions given, respectively, by:
 $MT_2 = T_1(\vec{v}) + T_2(\vec{v}),$
 $(T_1 - T_2)(\vec{v}) = T_1(\vec{v}) - T_2(\vec{v}),$ and
 $MT_2 = (kT_1)(\vec{v}) = kT_1(\vec{v}).$

Arithmetic of Matrices

Definitions: If A and B are both
$$m \times n$$
 matrices, and k is any scalar, then we can define:
 $A + B, A - B, \text{ and } kA$
as $m \times n$ matrices with entries given by:
Note: Matrix
Add/Sub need
them to be SAME
 $(A - B)_{i,j} = (A)_{i,j} + (B)_{i,j}, \quad what is also in leach is i$

We call these the sum and difference of *A* and *B*, and the scalar multiple of *A* by *k*.

In particular, we can define the *negative* of a matrix, -A, to be:

$$-A = (-1)A = \left(-\alpha_{ij}\right)_{m \times r}$$

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with the property that:

$$A + (-A) = (-A) + A = 0_{m \times n}$$

$$A + (-A) = [0]_{mxn}$$

Connection Between Linear Transformations and Matrices

Theorem: If $T_1 : \mathbb{R}^n \to \mathbb{R}^m$ and $T_2 : \mathbb{R}^n \to \mathbb{R}^m$ are linear transformations, with matrices $[T_1]$ and $[T_2]$ respectively, and k is any scalar, then for any $\vec{v} \in \mathbb{R}^n$:

$$(T_1 + T_2)(\vec{v}) = ([T_1] + [T_2])\vec{v}$$

$$(T_1 - T_2)(\vec{v}) = ([T_1] - [T_2])\vec{v}$$

$$(kT_1)(\vec{v}) = (k[T_1])\vec{v}$$

Consequently, $T_1 + T_2$, $T_1 - T_2$ and kT_1 are linear transformations with matrices given by, respectively:

$$[T_1 + T_2] = [T_1] + [T_2],$$

$$[T_1 - T_2] = [T_1] - [T_2], \text{ and}$$

$$[kT_1] = k[T_1]$$

Compositions of Linear Transformations

Definition/Theorem: If $T_1 : \mathbb{R}^n \to \mathbb{R}^k$ and $T_2 : \mathbb{R}^k \to \mathbb{R}^m$ are linear transformations, then we can define their **composition**:

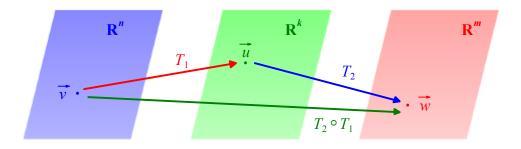
$$T_2 \circ T_1 : \mathbb{R}^n \to \mathbb{R}^m,$$

as a linear transformation, with action given as follows:

Suppose $\vec{u} \in \mathbb{R}^n$, $T_1(\vec{u}) = \vec{v} \in \mathbb{R}^k$, and $T_2(\vec{v}) = \vec{w} \in \mathbb{R}^m$. Then:

 $(T_2 \circ T_1)(\vec{u}) = T_2(T_1(\vec{u})) = T_2(\vec{v}) = \vec{w}.$

$$\mathbb{R}^{n} \xrightarrow{T_{1}} \mathbb{R}^{k} \xrightarrow{T_{2}} \mathbb{R}^{m}$$
$$\searrow T_{2} \circ T_{1} \nearrow$$



General Matrix Products

Definition — Matrix Product:

If A is an $m \times \underline{k}$ matrix, and B is a $\underline{k} \times n$, then we can construct the $m \times n$ matrix AB, where:

 $column \ i \ of \ AB = A \times (column \ i \ of \ B)$

In other words, if we write *B* in terms of its columns as:

$$B = \left[\vec{c}_1 \mid \vec{c}_2 \mid \cdots \mid \vec{c}_n\right]$$

then:

$$AB = \left[\vec{Ac_1} \mid \vec{Ac_2} \mid \cdots \mid \vec{Ac_n}\right]$$

NOTE: see the extra notes for an alternative way to think about matrix multiplication.

NOTE: entry in i-th row, j-th column is: (row i of A) o (column j of B) [yes, this is a dot product!)

Linear Combinations of Linear Transformations and Matrices

Definition of a linear combination of linear transformations T_1, T_2, ..., T_k.

$$(c_1 T_1 + c_2 T_2 + \dots + c_k T_k)(\vec{v})$$

= $c_1 T_1(\vec{v}) + c_2 T_2(\vec{v}) + \dots + c_k T_k(\vec{v})$

Note: for this to make sense, need all of them to be from same input same to same output space. Namely, T_i : $R^n --> R^m$ for all i=1,...,k

$c_1A_1+c_2A_2+\cdots+c_kA_k.$

Of course, by the Equivalence of LT and Mat Thm, anything we can do for LTs we can do for Mat.

$$[c_1T_1 + c_2T_2 + \dots + c_kT_k]$$

= $c_1[T_1] + c_2[T_2] + \dots + c_k[T_k]$

Not surprisingly, matrix of the LC of LTs equals the LC of Mats