

Task 1: Matrix representation of linear transformations

Let $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. If $\mathbf{e}_1, \dots, \mathbf{e}_n$ are the canonical basis vectors of \mathbb{R}^n , then the standard matrix of T is given by

$$S = [T(\mathbf{e}_1) \quad T(\mathbf{e}_2) \quad \cdots \quad T(\mathbf{e}_n)]$$

a) Find the standard matrix for the following linear transformation:

$$T(x_1, x_2) = \begin{bmatrix} 2x_1 + x_2 \\ x_1 - 3x_2 \\ -x_1 + x_2 \end{bmatrix}$$

b) Let $T_\theta \in \text{End}(\mathbb{R}^2)$ be the linear transformation that rotates $\mathbf{x} \in \mathbb{R}^2$ by the angle θ around the origin. Find the standard matrix of T .

Subtask a:

$$T(\mathbf{e}_1) = \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix} \quad T(\mathbf{e}_2) = \begin{bmatrix} 1 \\ -3 \\ 1 \end{bmatrix} \implies S = \begin{bmatrix} 2 & 1 \\ 1 & -3 \\ -1 & 1 \end{bmatrix}$$

Subtask b: Let $f: \mathbb{R}^2 \rightarrow \mathbb{C}$ with

$$f(\mathbf{x}) = f(x_1, x_2) = x_1 + ix_2$$

We see that f is bijective on \mathbb{R}^2 , hence \mathbb{C} and \mathbb{R}^2 are isomorphic under f with $f^{-1}(x_1 + ix_2) = \begin{bmatrix} x_1 & x_2 \end{bmatrix}^t$. Let $z \in \mathbb{C}$, then z has a polar-form $z = |z| \cdot e^{i\varphi}$. We want to show, that $f^{-1}(z)$ has angle θ to the x-axis. The standard-dot product in \mathbb{R}^n has the following property:

$$\langle \mathbf{x}, \mathbf{y} \rangle = \|\mathbf{x}\| \cdot \|\mathbf{y}\| \cdot \cos(\varphi)$$

We are interested in the angle of $f^{-1}(re^{i\theta})$ to \mathbf{e}_1 :

$$r \cos(\theta) = r \sqrt{\cos^2(\theta) + \sin^2(\theta)} \cos(\varphi) \iff \cos(\theta) = \cos(\varphi)$$

Since $\varphi \in [-\pi, \pi]$, we get $\theta = \varphi$. Notice that for $\zeta = e^{i\theta}$, we get:

$$z \cdot \zeta = |z|e^{i\varphi}e^{i\theta} = |z|e^{i(\varphi+\theta)}$$

Hence $z \in \mathbb{C}$: $|z| = 1$ act on \mathbb{C} like a rotation of θ around the origin. We need R_θ to be linear. Let $z_1, z_2 \in \mathbb{C}$ and $\lambda, \mu \in \mathbb{R}$, then:

$$R_\theta(\lambda z_1 + \mu z_2) = (\lambda z_1 + \mu z_2)e^{i\theta} = \lambda z_1 e^{i\theta} + \mu z_2 e^{i\theta} = \lambda R_\theta(z_1) + \mu R_\theta(z_2)$$

Now let $R_\theta(z) = ze^{i\theta}$, then $S_\theta = [f^{-1}(R_\theta(f(\mathbf{e}_1))) \quad f^{-1}(R_\theta(f(\mathbf{e}_2)))]$

$$\begin{aligned} f^{-1}(R_\theta(f(\mathbf{e}_1))) &= f^{-1}(e^{i\theta}) = \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix} & f^{-1}(R_\theta(f(\mathbf{e}_2))) &= f^{-1}(ie^{i\theta}) = \begin{bmatrix} -\sin(\theta) \\ \cos(\theta) \end{bmatrix} \\ \implies S_\theta &= \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \end{aligned}$$

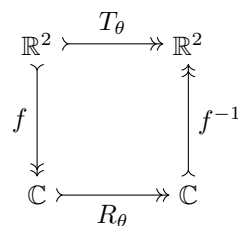


Figure 1: Visualization for finding S_θ

Task 2: Distances between points and linear manifolds

- a) Find the distance between the point $(-3, 1) = \mathbf{a}$ and the line $4x + 3y + 4 = 0$
b) Find the distance between the point $(3, 1, -2) = \mathbf{b}$ and the plane $x + 2y - 2z = 4$

We pursue a more general approach. Let $(V, \langle \cdot, \cdot \rangle)$ be an inner-product space, $\langle \cdot, \cdot \rangle : V^2 \rightarrow \mathbb{R}$, with the induced norm $\|\mathbf{x}\| = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$, and $U \subseteq V$ be a subspace of V . Let $\mathbf{x} \in V$, then we define the distance of \mathbf{x} to U as

$$d(\mathbf{x}, U) = \inf_{\mathbf{u} \in U} \|\mathbf{x} - \mathbf{u}\|$$

We prove the following:

$$d(\mathbf{x}, U) = \|\mathbf{x} - \pi_U(\mathbf{x})\|$$

Where π_U is the orthogonal projection of \mathbf{x} into U . Let $\mathbf{v} \in V$ and $\mathbf{u} \in U$, then

$$\begin{aligned} \|\mathbf{v} - \mathbf{u}\|^2 &= \langle \mathbf{v} - \mathbf{u}, \mathbf{v} - \mathbf{u} \rangle = \langle \mathbf{v} - \pi_U(\mathbf{v}) + \pi_U(\mathbf{v}) - \mathbf{u}, \mathbf{v} - \pi_U(\mathbf{v}) + \pi_U(\mathbf{v}) - \mathbf{u} \rangle \\ &= \langle \mathbf{v} - \pi_U(\mathbf{v}), \mathbf{v} - \pi_U(\mathbf{v}) \rangle + \langle \pi_U(\mathbf{v}) - \mathbf{u}, \pi_U(\mathbf{v}) - \mathbf{u} \rangle = \|\mathbf{v} - \pi_U(\mathbf{v})\|^2 + \underbrace{\|\mathbf{u} - \pi_U(\mathbf{v})\|^2}_{\geq 0} \geq \|\mathbf{v} - \pi_U(\mathbf{v})\|^2 \end{aligned}$$

Thus $\forall \mathbf{u} \in U: \|\mathbf{v} - \mathbf{u}\| \geq \|\mathbf{v} - \pi_U(\mathbf{v})\|$.

Let $(\mathbf{u}_1, \dots, \mathbf{u}_k)$ be a basis of U . Since the inner product in V is, at least, sesquilinear, π_U has the following properties:

$$\begin{aligned} \pi_U(\mathbf{x}) &= \sum_{i=1}^k c_i \mathbf{u}_i \\ \forall j = 1, \dots, k: \langle \pi_U(\mathbf{x}) - \mathbf{x}, \mathbf{u}_j \rangle &= 0 \iff \forall j = 1, \dots, k: \langle \pi_U(\mathbf{x}), \mathbf{u}_j \rangle = \langle \mathbf{x}, \mathbf{u}_j \rangle \end{aligned}$$

We get the following linear system of equations:

$$\begin{aligned} \langle \pi_U(\mathbf{x}), \mathbf{u}_j \rangle &= \langle \mathbf{x}, \mathbf{u}_j \rangle \iff \left\langle \sum_{i=1}^k c_i \mathbf{u}_i, \mathbf{u}_j \right\rangle = \langle \mathbf{x}, \mathbf{u}_j \rangle \\ \iff \sum_{i=1}^k c_i \langle \mathbf{u}_i, \mathbf{u}_j \rangle &= \langle \mathbf{x}, \mathbf{u}_j \rangle \end{aligned}$$

We denote

$$\text{Gram}(\mathbf{u}_1, \dots, \mathbf{u}_k) = \begin{bmatrix} \langle \mathbf{u}_1, \mathbf{u}_1 \rangle & \langle \mathbf{u}_1, \mathbf{u}_2 \rangle & \cdots & \cdots & \langle \mathbf{u}_1, \mathbf{u}_k \rangle \\ \langle \mathbf{u}_2, \mathbf{u}_1 \rangle & \langle \mathbf{u}_2, \mathbf{u}_2 \rangle & \ddots & & \vdots \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ \vdots & \vdots & & \ddots & \langle \mathbf{u}_{k-1}, \mathbf{u}_k \rangle \\ \langle \mathbf{u}_k, \mathbf{u}_1 \rangle & \langle \mathbf{u}_k, \mathbf{u}_2 \rangle & \cdots & \cdots & \langle \mathbf{u}_k, \mathbf{u}_k \rangle \end{bmatrix}$$

the Gramian-Matrix of $(\mathbf{u}_1, \dots, \mathbf{u}_k)$, then

$$\text{Gram}(\mathbf{u}_1, \dots, \mathbf{u}_k) \mathbf{c} = \begin{bmatrix} \langle \mathbf{x}, \mathbf{u}_1 \rangle \\ \vdots \\ \langle \mathbf{x}, \mathbf{u}_k \rangle \end{bmatrix} \quad \mathbf{c} = \begin{bmatrix} c_1 \\ \vdots \\ c_k \end{bmatrix}$$

Let $u_{ij} = \langle \mathbf{u}_i, \mathbf{u}_j \rangle$. By the properties of $\langle \cdot, \cdot \rangle$, we know $u_{ij} = \overline{u_{ji}}$, thus $\text{Gram}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ is hermitian. Let $\mathbf{U} = [\mathbf{u}_1^t \cdots \mathbf{u}_k^t]^t$. If $\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{u}^t \mathbf{P} \mathbf{v}$, then

$$\text{Gram}(\mathbf{u}_1, \dots, \mathbf{u}_k) = \mathbf{U}^t \mathbf{P} \mathbf{U}$$

Let $\mathbf{x} \in V \setminus \{\mathbf{0}\}$: $U\mathbf{x} = \mathbf{0}$:

$$\mathbf{0} = U\mathbf{x} \iff \mathbf{0} = U^t P U \mathbf{x} = \text{Gram}(\mathbf{u}_1, \dots, \mathbf{u}_k) \mathbf{x}$$

Notice that $U\mathbf{x}$ can only be $\mathbf{0}$ for $\mathbf{x} \neq \mathbf{0}$, if the columns of U are linearly dependent. By requirement $\mathbf{u}_1, \dots, \mathbf{u}_k$ form a basis of U and are thus linearly independent, thusly $\text{Gram}(\mathbf{u}_1, \dots, \mathbf{u}_k)$ is regular.

If $M = \mathbf{u} + U$ is a linear manifold, we apply the inverse translation $T_{\mathbf{u}}^{-1}(\mathbf{x}) = \mathbf{x} - \mathbf{u}$ to make $T_{\mathbf{u}}^{-1}[M] = U$ a (linear) subspace of V . Notice that $T_{\mathbf{u}}^{-1}$ is an isometry, i.e. the following holds:

$$\begin{aligned} \|T_{\mathbf{u}}(\mathbf{x}) - T_{\mathbf{u}}(\mathbf{y})\| &= \|\mathbf{x} + \mathbf{u} - \mathbf{y} - \mathbf{u}\| = \|\mathbf{x} - \mathbf{y}\| \\ d(T_{\mathbf{u}}(\mathbf{x}), M) &= \|T_{\mathbf{u}}(\mathbf{x}) - \pi_M(\mathbf{x})\| = \|T_{\mathbf{u}}(\mathbf{x}) - T_{\mathbf{u}}(\pi_U(\mathbf{x}))\| = \|\mathbf{x} - \pi_U(\mathbf{x})\| = d(\mathbf{x}, U) \end{aligned}$$

Subtask a: We first need to find a subspace \tilde{G} and $\mathbf{u} \in \mathbb{R}^2$, such that $G = \{(x, y) \in \mathbb{R}^2 : 4x + 3y + 4 = 0\} = \mathbf{u} + \tilde{G}$. We solve for y :

$$y = -\frac{4}{3}x - \frac{4}{3}$$

Thus:

$$G = \left\{ \lambda \begin{bmatrix} 3 \\ -4 \end{bmatrix} + \begin{bmatrix} 0 \\ -\frac{4}{3} \end{bmatrix} \mid \lambda \in \mathbb{R} \right\} = \text{span} \left(\begin{bmatrix} 3 \\ -4 \end{bmatrix} \right) + \begin{bmatrix} 0 \\ -\frac{4}{3} \end{bmatrix}$$

Notice $\text{Gram}(\mathbf{u}_1) = \|\mathbf{u}_1\|^2 = 25$ and therefore

$$\pi_{\tilde{G}}(\mathbf{x}) = \frac{1}{25} \langle \mathbf{x}, \mathbf{u}_1 \rangle \mathbf{u}_1$$

Now:

$$\begin{aligned} T_{\mathbf{u}}^{-1}(\mathbf{x}) &= \mathbf{x} + \begin{bmatrix} 0 \\ \frac{4}{3} \end{bmatrix} \implies T_{\mathbf{u}}^{-1}(\mathbf{a}) = \begin{bmatrix} -3 \\ \frac{7}{3} \end{bmatrix} \implies \pi_{\tilde{G}}(T_{\mathbf{u}}^{-1}(\mathbf{a})) = \frac{1}{25} \left(-9 - \frac{28}{3} \right) \begin{bmatrix} 3 \\ -4 \end{bmatrix} \\ &= \frac{55}{25} \frac{1}{3} \begin{bmatrix} -3 \\ 4 \end{bmatrix} = \frac{11}{15} \begin{bmatrix} -3 \\ 4 \end{bmatrix} \\ \implies d(T_{\mathbf{u}}^{-1}(\mathbf{a}), \tilde{G}) &= \left\| \frac{11}{15} \begin{bmatrix} -3 \\ 4 \end{bmatrix} - \begin{bmatrix} -3 \\ \frac{7}{3} \end{bmatrix} \right\| = \left\| \begin{bmatrix} \frac{4}{5} \\ \frac{3}{5} \end{bmatrix} \right\| = \frac{1}{5} \|[3 \quad 4]^t\| = 1 \end{aligned}$$

Subtask b: We again need to find a subspace \tilde{E} and a vector \mathbf{u} , such that

$$E = \{(x, y, z) \in \mathbb{R}^3 : x + 2y - 2z = 4\} = \mathbf{u} + \tilde{E}$$

We solve for z :

$$z = \frac{1}{2}x + y - 2 \implies E = \left\{ \lambda \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} + \mu \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ -2 \end{bmatrix} \mid \lambda, \mu \in \mathbb{R} \right\} = \text{span} \left(\begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right) + \begin{bmatrix} 0 \\ 0 \\ -2 \end{bmatrix}$$

We compute $\text{Gram}(\mathbf{u}_1, \mathbf{u}_2)$ and it's inverse:

$$\text{Gram}(\mathbf{u}_1, \mathbf{u}_2) = \begin{bmatrix} 5 & 1 \\ 1 & 2 \end{bmatrix} \quad \text{Gram}^{-1}(\mathbf{u}_1, \mathbf{u}_2) = \frac{1}{9} \begin{bmatrix} 2 & -1 \\ -1 & 5 \end{bmatrix}$$

Now:

$$\begin{aligned} T_{\mathbf{u}}^{-1}(\mathbf{b}) &= \begin{bmatrix} 3 \\ 1 \\ 0 \end{bmatrix} \quad \langle T_{\mathbf{u}}^{-1}(\mathbf{b}), \mathbf{u}_1 \rangle = 6 \quad \langle T_{\mathbf{u}}^{-1}(\mathbf{b}), \mathbf{u}_2 \rangle = 1 \\ \mathbf{c} &= \frac{1}{9} \begin{bmatrix} 2 & -1 \\ -1 & 5 \end{bmatrix} \begin{bmatrix} 6 \\ 1 \end{bmatrix} = \frac{1}{9} \begin{bmatrix} 11 \\ -1 \end{bmatrix} \implies \pi_{\tilde{E}}(T_{\mathbf{u}}^{-1}(\mathbf{b})) = \frac{11}{9} \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} - \frac{1}{9} \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{9} \begin{bmatrix} 22 \\ -1 \\ 10 \end{bmatrix} \\ \implies d(T_{\mathbf{u}}^{-1}(\mathbf{b}), \tilde{E}) &= \left\| \frac{1}{9} \begin{bmatrix} 22 \\ -1 \\ 10 \end{bmatrix} - \begin{bmatrix} 3 \\ 1 \\ 0 \end{bmatrix} \right\| = \frac{1}{9} \left\| \begin{bmatrix} -5 \\ -10 \\ 10 \end{bmatrix} \right\| = \frac{1}{9} \sqrt{225} = \frac{15}{9} = \frac{5}{3} \end{aligned}$$

Task 3: Matrix Rank

We are given the matrix

$$A(r, s) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & r-2 & 2 \\ 0 & s-1 & r+2 \\ 0 & 0 & 3 \end{bmatrix}$$

Do there exist values for r, s , such that $\text{rank}A(r, s) = 1$ and $\text{rank}A(r, s) = 2$ respectively? If so, find them.

Since the first row $r_1 = [1 \ 0 \ 0]$ is linearly independent from the remaining three rows, we get that $\forall r, s \in \mathbb{R}: \text{rank}A(r, s) \geq 1$. Let

$$M(r, s) = \begin{bmatrix} r-2 & 2 \\ s-1 & r+2 \end{bmatrix}$$

First let's analyze some special cases. If $r = 2$, then

$$\text{rank}M(2, s) = \begin{cases} 1 & s = 1 \\ 2 & s \neq 1 \end{cases}$$

If $r = -2$, then

$$\text{rank}M(-2, s) = \begin{cases} 1 & s = 1 \\ 2 & s \neq 1 \end{cases}$$

If $\text{rank}M(r, s) = 2$, then we can bring M into diagonal-form using elementary row-operations. Thus we can eliminate the fourth row $r_4 = [0 \ 0 \ 3]$ of $A(r, s)$, and hence $\text{rank}A(r, s) = 3$.

$$\begin{bmatrix} r-2 & 2 \\ s-1 & r+2 \end{bmatrix} \xrightarrow{\frac{1}{r-2}I, \frac{1}{s-1}II} \begin{bmatrix} 1 & \frac{2}{r-2} \\ 1 & \frac{r+2}{s-1} \end{bmatrix} \xrightarrow{II-I} \begin{bmatrix} 1 & \frac{2}{r-2} \\ 0 & \frac{r+2}{s-1} - \frac{2}{r-2} \end{bmatrix}$$

We see that $\text{rank}M(r, s) = 1$ iff

$$\begin{aligned} \frac{r+2}{s-1} - \frac{2}{r-2} &= 0 \\ \frac{r+2}{s-1} - \frac{2}{r-2} &= \frac{r^2 - 4 - 2s + 2}{(s-1)(r-2)} = \frac{r^2 - 2s - 2}{(s-1)(r-2)} \stackrel{!}{=} 0 \implies r^2 - 2s - 2 = 0 \\ r &= \pm\sqrt{2}\sqrt{s+1} \quad s = \frac{r^2}{2} - 1 \end{aligned}$$

Thus we get:

$$\text{rank}M(r, s) = \begin{cases} 1 & r = \pm 2 \wedge s = 1 \\ 1 & r = \pm\sqrt{2s+2} \wedge s \in \mathbb{R} \setminus \{1\} \\ 1 & s = \frac{r^2}{2} - 1 \wedge r \in \mathbb{R} \setminus \{2\} \\ 2 & \text{else} \end{cases}$$

Notice

$$M(1, 2) = \begin{bmatrix} 0 & 2 \\ 0 & 4 \end{bmatrix} \rightsquigarrow \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

Thus for $r = 2$ and $s = 1$, we get $\text{rank}A(2, 1) = 2$, otherwise $\text{rank}A(r, s)$ is 3. Hence there does not exist any pair $r, s \in \mathbb{R}$, such that $\text{rank}A(r, s) = 1$.

Task 4: Matrix Transformation

Let $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a mapping of the form

$$T_A(\mathbf{x}) = A\mathbf{x}$$

where $\mathbf{x} \in \mathbb{R}^n$ and $A \in \mathbb{R}^{m \times n}$. Let $T_A, T_B: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be two matrix transformations. Prove the following

$$\forall \mathbf{x} \in \mathbb{R}^n: T_A(\mathbf{x}) = T_B(\mathbf{x}) \implies A = B$$

Notice that a linear transformation is uniquely defined by it's image of the basis-vectors. Let $\mathbf{x} \in V$ and $T \in \text{Hom}(V, W)$, V and W vector-spaces, then

$$T(\mathbf{x}) = T\left(\sum_{i=1}^n \lambda_i \mathbf{b}_i\right) = \sum_{i=1}^n \lambda_i T(\mathbf{b}_i)$$

Now:

$$A = S_A = [T_A(\mathbf{b}_1) \quad \cdots \quad T_A(\mathbf{b}_n)] = [T_B(\mathbf{b}_1) \quad \cdots \quad T_B(\mathbf{b}_n)] = S_B = B$$

Task 5: Properties of Euclidean Norm

Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n \setminus \{\mathbf{0}\}$, prove the following holds:

$$\left\| \frac{\mathbf{x}}{\|\mathbf{x}\|^2} - \frac{\mathbf{y}}{\|\mathbf{y}\|^2} \right\| = \frac{\|\mathbf{x} - \mathbf{y}\|}{\|\mathbf{x}\| \cdot \|\mathbf{y}\|}$$

$$\begin{aligned} \left(\frac{\|\mathbf{x} - \mathbf{y}\|}{\|\mathbf{x}\| \cdot \|\mathbf{y}\|} \right)^2 &= \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle \frac{1}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2} = \frac{\langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle - \langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2} \\ &= \frac{\langle \mathbf{x}, \mathbf{x} \rangle - \langle \mathbf{x}, \mathbf{y} \rangle - \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2} = \frac{\|\mathbf{x}\|^2 + \|\mathbf{y}\|^2 - 2\langle \mathbf{x}, \mathbf{y} \rangle}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2} \end{aligned}$$

And:

$$\begin{aligned} \left(\left\| \frac{\mathbf{x}}{\|\mathbf{x}\|^2} - \frac{\mathbf{y}}{\|\mathbf{y}\|^2} \right\| \right)^2 &= \left\langle \frac{\mathbf{x}}{\|\mathbf{x}\|^2} - \frac{\mathbf{y}}{\|\mathbf{y}\|^2}, \frac{\mathbf{x}}{\|\mathbf{x}\|^2} - \frac{\mathbf{y}}{\|\mathbf{y}\|^2} \right\rangle = \left\langle \frac{\|\mathbf{y}\|^2 \mathbf{x} - \|\mathbf{x}\|^2 \mathbf{y}}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2}, \frac{\|\mathbf{y}\|^2 \mathbf{x} - \|\mathbf{x}\|^2 \mathbf{y}}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2} \right\rangle \\ &= \frac{1}{\|\mathbf{x}\|^4 \cdot \|\mathbf{y}\|^4} \langle \|\mathbf{y}\|^2 \mathbf{x} - \|\mathbf{x}\|^2 \mathbf{y}, \|\mathbf{y}\|^2 \mathbf{x} - \|\mathbf{x}\|^2 \mathbf{y} \rangle = \frac{\|\mathbf{y}\|^2 \langle \mathbf{x}, \|\mathbf{y}\|^2 \mathbf{x} - \|\mathbf{x}\|^2 \mathbf{y} \rangle - \|\mathbf{x}\|^2 \langle \mathbf{y}, \|\mathbf{y}\|^2 \mathbf{x} - \|\mathbf{x}\|^2 \mathbf{y} \rangle}{\|\mathbf{x}\|^4 \cdot \|\mathbf{y}\|^4} \\ &= \frac{\|\mathbf{y}\|^2 \langle \mathbf{x}, \|\mathbf{y}\|^2 \mathbf{x} \rangle - \|\mathbf{y}\|^2 \langle \mathbf{x}, \|\mathbf{x}\|^2 \mathbf{y} \rangle - \|\mathbf{x}\|^2 \langle \mathbf{y}, \|\mathbf{y}\|^2 \mathbf{x} \rangle + \|\mathbf{x}\|^2 \langle \mathbf{y}, \|\mathbf{y}\|^2 \mathbf{y} \rangle}{\|\mathbf{y}\|^4 \cdot \|\mathbf{y}\|^4} \\ &= \frac{\|\mathbf{y}\|^4 \cdot \|\mathbf{x}\|^2 - \|\mathbf{y}\|^2 \cdot \|\mathbf{x}\|^2 \langle \mathbf{x}, \mathbf{y} \rangle - \|\mathbf{y}\|^2 \cdot \|\mathbf{x}\|^2 \langle \mathbf{y}, \mathbf{x} \rangle + \|\mathbf{x}\|^4 \cdot \|\mathbf{y}\|^2}{\|\mathbf{x}\|^4 \cdot \|\mathbf{y}\|^4} \\ &= \frac{\|\mathbf{y}\|^4 \cdot \|\mathbf{x}\|^2 - 2\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2 \langle \mathbf{x}, \mathbf{y} \rangle + \|\mathbf{x}\|^4 \cdot \|\mathbf{y}\|^2}{\|\mathbf{x}\|^4 \cdot \|\mathbf{y}\|^4} = \frac{\|\mathbf{y}\|^2 - 2\langle \mathbf{x}, \mathbf{y} \rangle + \|\mathbf{x}\|^2}{\|\mathbf{x}\|^2 \cdot \|\mathbf{y}\|^2} \end{aligned}$$

Task 6: Linear Independence

Let V be a \mathbb{K} vector-space and $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$, where $k \geq 2$. Prove that $(\mathbf{v}_1, \dots, \mathbf{v}_k)$ is linearly independent iff

$$\mathbf{v}_1 \neq \mathbf{0} \wedge \forall i = 1, \dots, k-1: \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_i) \subset \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_{i+1})$$

We use contraposition. If $(\mathbf{v}_1, \dots, \mathbf{v}_k)$ is linearly dependent, then

$$\mathbf{v}_1 = \mathbf{0} \vee \exists i \in \{1, \dots, k-1\}: \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_i) = \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_{i+1})$$

Assume $\mathbf{v}_1 = \mathbf{0}$, then

$$\sum_{i=1}^k \lambda_i \mathbf{v}_i = \lambda_1 \mathbf{v}_1 + \mathbf{0} = \mathbf{0} \forall \lambda_1 \in \mathbb{R}$$

Thus $(\mathbf{v}_1, \dots, \mathbf{v}_k)$ is linearly dependent. Now let $\mathbf{v}_1 \neq \mathbf{0}$. Assume $\exists i \in \{1, \dots, k-1\}$ such that $(\mathbf{v}_1, \dots, \mathbf{v}_i)$ is linearly independent and:

$$U_i = \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_i) = \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_{i+1})$$

Let $\mathbf{x} \in U_i \setminus \{\mathbf{0}\}$ and $\mu_{i+1} \neq 0$:

$$\begin{aligned} \mathbf{x} &= \sum_{l=1}^i \lambda_l \mathbf{v}_l = \sum_{l=1}^{i+1} \mu_l \mathbf{v}_l = \mu_{i+1} \mathbf{v}_{i+1} + \sum_{l=1}^i \mu_l \mathbf{v}_l \\ \iff \mathbf{0} &= \mu_{i+1} \mathbf{v}_{i+1} + \sum_{l=1}^i (\mu_l - \lambda_l) \mathbf{v}_l \iff \mu_{i+1} \mathbf{v}_{i+1} = \sum_{l=1}^i (\lambda_l - \mu_l) \mathbf{v}_l \\ \iff \mathbf{v}_{i+1} &= \sum_{l=1}^i \frac{\lambda_l - \mu_l}{\mu_{i+1}} \mathbf{v}_l \end{aligned}$$

We found a linear combination for \mathbf{v}_{i+1} in U_i , thus $(\mathbf{v}_1, \dots, \mathbf{v}_{i+1})$ is not linearly independent, thus $(\mathbf{v}_1, \dots, \mathbf{v}_k)$ is linearly dependent.