

Reference Sheet for CO233 Computational Methods

Autumn 2017

1 The \mathbb{R}^n and \mathbb{C}^n Vector Spaces

- *Convex combination*: triangle $a\mathbf{u} + b\mathbf{v} + c\mathbf{w}$ with vertices $\mathbf{u}, \mathbf{v}, \mathbf{w}$ and $0 \leq a, b, c \leq 1$ and $a + b + c = 1$, for $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{R}^3$.
- *Inner product*: $\mathbf{u} \cdot \mathbf{v} = \sum_{i=1}^3 \mathbf{u}_i \mathbf{v}_i = \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta$ for $\mathbf{u}, \mathbf{v} \in \mathbb{R}^3$, where $\|\mathbf{u}\| = \sqrt{\mathbf{u}_1^2 + \mathbf{u}_2^2 + \mathbf{u}_3^2}$.
- *Inner product*: $\langle \mathbf{v}, \mathbf{w} \rangle = \sum_{i=1}^n \mathbf{v}_i^* \mathbf{w}_i$ for $\mathbf{v}, \mathbf{w} \in \mathbb{C}^n$.
- *Norm*: $\|\mathbf{v}\| = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}$.

Representation of Linear Maps

- *Linear map* $\mathbb{R}^n \rightarrow \mathbb{R}^m$: can be represented by the real matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$:
 - For all $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$, $\mathbf{A}(\mathbf{v} + \mathbf{w}) = \mathbf{A}\mathbf{v} + \mathbf{A}\mathbf{w}$.
 - For all $\mathbf{v} \in \mathbb{R}^n, c \in \mathbb{R}$, $\mathbf{A}(c\mathbf{v}) = c(\mathbf{A}\mathbf{v})$.
- Extends simply to \mathbb{C}^n .

2 Norms

2.1 Vector Norms

A vector norm on \mathbb{R}^n is a real-valued map

$$\|\cdot\| : \mathbb{R}^n \rightarrow \mathbb{R}$$

which satisfies:

1. For any non-zero vector $\mathbf{x} \in \mathbb{R}^n$, $\|\mathbf{x}\| > 0$.
2. For any scalar λ and $\mathbf{x} \in \mathbb{R}^n$, $\|\lambda\mathbf{x}\| = |\lambda| \|\mathbf{x}\|$
3. For two vectors $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, $\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\|$.

The l_p Norms

$$\|\mathbf{v}\|_p = \left(\sum_{i=1}^n |\mathbf{v}_i|^p \right)^{1/p}$$

Properties

- For any vector \mathbf{x} , $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1$.

2.2 Cauchy-Schwartz Inequality

$$|\langle \mathbf{u}, \mathbf{v} \rangle| \leq \|\mathbf{u}\| \|\mathbf{v}\|$$

Proof

- Consider $\lambda\mathbf{u} + \mathbf{v}$.
- Since the length of any vector is non-negative, $0 \leq (\lambda\mathbf{u} + \mathbf{v}) \cdot (\lambda\mathbf{u} + \mathbf{v})$.
- Therefore $\lambda = \frac{\langle \mathbf{u}, \mathbf{v} \rangle}{\|\mathbf{u}\|^2}$.

2.3 Matrix Norms

A matrix norm on $\mathbb{R}^{m \times n}$ is a real-valued map.

3 Linear Independence

For $\mathbf{a}_i \in \mathbb{R}^m$, with $i = 1, \dots, k$, the \mathbf{a}_i s are linearly independent if whenever $\mathbf{x}_i \in \mathbb{R}$, we have $\sum_{i=1}^k \mathbf{x}_i \mathbf{a}_i = 0$, then $\mathbf{x}_i = 0$ for $i = 1, \dots, k$.

Methods for Determining Linear Independence Columns of \mathbf{A} are linearly independent if:

- Calculate determinant. Find $\det(\mathbf{A}) \neq 0$.
- Solve $\mathbf{A}\mathbf{x} = \mathbf{0}$. Find $\mathbf{x} = \mathbf{0}$.