

AMATH 352: Problem Set 6

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Norms and Inner products:

1. Let \mathbf{W} be an invertible matrix. Show that the map

$$\|\mathbf{x}\|_{\mathbf{W}} = \|\mathbf{W}\mathbf{x}\|_2$$

is a norm on \mathbb{R}^m . Is it still a norm if \mathbf{W} is singular? Why or why not?

Solution:

If \mathbf{W} is invertible, $\|\cdot\|_{\mathbf{W}}$ satisfies the 4 properties of a norm:

- (a) $\|\mathbf{x}\|_{\mathbf{W}} = \|\mathbf{W}\mathbf{x}\|_2 \geq 0 \ \forall \ \mathbf{x} \neq \mathbf{0}$ because $\|\cdot\|_2$ is a norm on \mathbb{R}^m , and $\text{null}(\mathbf{W}) = \{\mathbf{0}\}$.
- (b) Since \mathbf{W} is invertible, $\mathbf{W}\mathbf{x} = \mathbf{0} \implies \mathbf{x} = \mathbf{0}$, so $\|\mathbf{x}\|_{\mathbf{W}} = \|\mathbf{W}\mathbf{x}\|_2 = \mathbf{0} \implies \mathbf{x} = \mathbf{0}$.
- (c)

$$\begin{aligned}\|\alpha\mathbf{x}\|_{\mathbf{W}} &= \|\mathbf{W}(\alpha\mathbf{x})\|_2 \\ &= \|\alpha\mathbf{W}\mathbf{x}\|_2 \\ &= |\alpha| \|\mathbf{W}\mathbf{x}\|_2 \\ &= |\alpha| \|\mathbf{x}\|_{\mathbf{W}}\end{aligned}$$

- (d) By distributing the multiplication by \mathbf{W} , we see that all of the following statements are equivalent:

$$\begin{aligned}\|\mathbf{x} + \mathbf{y}\|_{\mathbf{W}} &\leq \|\mathbf{x}\|_{\mathbf{W}} + \|\mathbf{y}\|_{\mathbf{W}} \\ \|\mathbf{W}(\mathbf{x} + \mathbf{y})\|_2 &\leq \|\mathbf{W}\mathbf{x}\|_2 + \|\mathbf{W}\mathbf{y}\|_2 \\ \|\mathbf{W}\mathbf{x} + \mathbf{W}\mathbf{y}\|_2 &\leq \|\mathbf{W}\mathbf{x}\|_2 + \|\mathbf{W}\mathbf{y}\|_2\end{aligned}$$

We know the last version is satisfied because $\|\cdot\|_2$ satisfies the triangle inequality.

If \mathbf{W} is singular, then $\|\cdot\|_{\mathbf{W}}$ is *not* a norm; by selecting a non-zero $\mathbf{x} \in \text{null}(\mathbf{W})$, we can violate the first, second, or third properties of a norm.

2. Consider a real square matrix $\mathbf{M} \in \mathbb{R}^{n \times n}$. Suppose \mathbf{M} is symmetric and full rank. Furthermore, suppose \mathbf{M} is positive definite, i.e. it satisfies

$$\forall \mathbf{x} \in \mathbb{R}^n, \mathbf{x} \neq \mathbf{0} \implies \mathbf{x}^T \mathbf{M} \mathbf{x} > 0.$$

Show that the map $\langle \cdot, \cdot \rangle_{\mathbf{M}} : \mathbb{R}^n \rightarrow \mathbb{R}$ given by

$$\langle \mathbf{x}, \mathbf{y} \rangle_{\mathbf{M}} = \mathbf{x}^T \mathbf{M} \mathbf{y}$$

is an inner product on \mathbb{R}^n .

Solution:

Checking the five properties of an inner product:

- (a) $\langle \mathbf{x}, \mathbf{x} \rangle_{\mathbf{M}} \geq 0 \ \forall \ \mathbf{x}$ is true by the definition of a positive definite matrix.
- (b) $\langle \mathbf{x}, \mathbf{x} \rangle_{\mathbf{M}} = 0 \implies \mathbf{x} = \mathbf{0}$ is also true because \mathbf{M} is a square matrix of full rank, $\text{null}(\mathbf{M}) = \{\mathbf{0}\}$.
- (c)

$$\begin{aligned}\langle \alpha \mathbf{x}, \mathbf{y} \rangle_{\mathbf{M}} &= (\alpha \mathbf{x})^T \mathbf{M} \mathbf{y} \\ &= \alpha (\mathbf{x}^T \mathbf{M} \mathbf{y}) \\ &= \alpha \langle \mathbf{x}, \mathbf{y} \rangle_{\mathbf{M}}\end{aligned}$$

(d)

$$\begin{aligned}\langle \mathbf{x}, \mathbf{y} \rangle_{\mathbf{M}} &= \mathbf{x}^T \mathbf{M} \mathbf{y} \\ &= \mathbf{y}^T \mathbf{M}^T \mathbf{x} \\ &= \mathbf{y}^T \mathbf{M} \mathbf{x} \\ &= \langle \mathbf{y}, \mathbf{x} \rangle_{\mathbf{M}}\end{aligned}$$

Since $\langle \cdot, \cdot \rangle_{\mathbf{M}} \in \mathbb{R}$, and a scalar is equal to its transpose, we rewrite the expression in step 2 using the rule of matrix products and the transpose operator. Since \mathbf{M} is symmetric, $\mathbf{M}^T = \mathbf{M}$ in the third step.

(e)

$$\begin{aligned}\langle \mathbf{x} + \mathbf{z}, \mathbf{y} \rangle_{\mathbf{M}} &= (\mathbf{x} + \mathbf{z})^T \mathbf{M} \mathbf{y} \\ &= \mathbf{x}^T \mathbf{M} \mathbf{y} + \mathbf{z}^T \mathbf{M} \mathbf{y} \\ &= \langle \mathbf{x}, \mathbf{y} \rangle_{\mathbf{M}} + \langle \mathbf{z}, \mathbf{y} \rangle_{\mathbf{M}}\end{aligned}$$

$\langle \cdot, \cdot \rangle_{\mathbf{M}} : \mathbb{R}^n \rightarrow \mathbb{R}$ therefore satisfies all five properties required of an inner product.

3. Show that the function $\langle \cdot, \cdot \rangle : C^0([-1, 1], \mathbb{R}) \rightarrow \mathbb{R}$ given by

$$\langle f, g \rangle = \int_{-1}^1 f(x)g(x)dx$$

is an inner product on $C^0([-1, 1], \mathbb{R})$. Note that $C^0([-1, 1], \mathbb{R})$ denotes the space of continuous functions that take input from $[-1, 1]$ and produce output in \mathbb{R} .

Solution:

Checking the five properties of an inner product:

- (a) $\forall f \neq 0, \langle f, f \rangle \geq 0$, since the product $f(x)f(x) = f^2(x)$ is always non-negative. ($f(x) \in \mathbb{R}$.)
- (b) From the above we know that $\langle f, f \rangle = 0 \implies f(x) = 0$, (which is the zero element for this operator) since if f were anywhere $\neq 0$ on the interval $[-1, 1]$, the integral would be positive.
- (c)

$$\begin{aligned}\langle \alpha f, g \rangle &= \int_{-1}^1 \alpha f(x)g(x)dx \\ &= \alpha \int_{-1}^1 f(x)g(x)dx \\ &= \alpha \langle f, g \rangle\end{aligned}$$

(d)

$$\begin{aligned}\langle f, g \rangle &= \int_{-1}^1 f(x)g(x)dx \\ &= \int_{-1}^1 g(x)f(x)dx \\ &= \langle g, f \rangle\end{aligned}$$

(e)

$$\begin{aligned}\langle f + h, g \rangle &= \int_{-1}^1 (f(x) + h(x))g(x)dx \\ &= \int_{-1}^1 (f(x)g(x) + h(x)g(x))dx \\ &= \int_{-1}^1 f(x)g(x)dx + \int_{-1}^1 h(x)g(x)dx \\ &= \langle f, g \rangle + \langle h, g \rangle\end{aligned}$$

$\langle f, g \rangle$ therefore satisfies all five properties required of an inner product.

Conditioning:

4. In this problem we show that orthogonal matrices are “perfectly conditioned” in the sense that their condition numbers with respect to the 2-norm are always 1. Suppose $\mathbf{O} \in \mathbb{R}^{n \times n}$ is an orthogonal matrix.

- (a) Show that $\|\mathbf{O}\|_2 = 1$. (Hint: recall that orthogonal matrices preserve the 2-norms of vectors).
- (b) Show that \mathbf{O}^T is an orthogonal matrix.
- (c) Show that $\kappa_2(\mathbf{O}) = 1$.

Solution:

- (a) $\|\mathbf{O}\|_2 = \max \frac{\|\mathbf{O}\mathbf{x}\|_2}{\|\mathbf{x}\|_2}$, but since \mathbf{O} is orthogonal, we know that $\|\mathbf{O}\mathbf{x}\|_2 = \|\mathbf{x}\|_2$ for all \mathbf{x} . So this reduces to $\|\mathbf{O}\|_2 = \max \frac{\|\mathbf{O}\mathbf{x}\|_2}{\|\mathbf{x}\|_2} = \max \frac{\|\mathbf{x}\|_2}{\|\mathbf{x}\|_2} = 1$.
- (b) By definition, an orthogonal matrix is a square matrix with $\mathbf{O}^T = \mathbf{O}^{-1}$. By inverting both sides, we can see that $(\mathbf{O}^T)^{-1} = \mathbf{O}$. Since the inverse of \mathbf{O}^T is equal to its transpose, \mathbf{O}^T is orthogonal.
- (c) $\kappa_2(\mathbf{O}) = \|\mathbf{O}\|_2 \|\mathbf{O}^{-1}\|_2$, but by definition, $\mathbf{O}^{-1} = \mathbf{O}^T$, so this is equivalent to $\|\mathbf{O}\|_2 \|\mathbf{O}^T\|_2$. From the previous problems, we know that \mathbf{O}^T is orthogonal and that the 2-norm of any orthogonal matrix is 1. So this reduces to

$$\kappa_2(\mathbf{O}) = \|\mathbf{O}\|_2 \|\mathbf{O}^T\|_2 = 1$$

Operation count:

5. Find the number of necessary floating point operations required to compute the following operations (using the big-oh notation introduced in class). Explain your reasoning in each case.

- (a) Compute the sum $\mathbf{A} + \mathbf{B}$ for $\mathbf{A}, \mathbf{B} \in \mathbb{R}^{m \times n}$.
- (b) Compute the outer product $\mathbf{u}\mathbf{v}^T$ for $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$.

- (c) Compute the product $\mathbf{A}\mathbf{x}$ for $\mathbf{x} \in \mathbb{R}^n$ and $\mathbf{A} \in \mathbb{R}^{n \times n}$ where \mathbf{A} is upper triangular.

Solution:

- (a) Each element of the sum requires a single addition, and there are $m \times n$ elements in $\mathbf{A} + \mathbf{B} \in \mathbb{R}^{m \times n}$, so this is $O(mn)$.
- (b) $\mathbf{u}\mathbf{v}^T \in \mathbb{R}^{n \times n}$, and each element of the product requires a single multiplication, so the total operation count is $O(n^2)$.
- (c) This is equivalent to n dot-products of n -dimensional vectors. Each dot-product requires n multiplies and $n - 1$ additions, or $2n - 1$ total flops. So the total operation count is $2n^2 - n$, or $O(n^2)$.

Note that because \mathbf{A} is upper-triangular, only half of the operations will be with non-zero components, but the resulting reduction in flops of $\frac{1}{2}$ is irrelevant for big-O.

6. **Comparing growth rates:** This exercise is meant to give you an idea of how quickly the number of flops required to solve a problem can increase when one increases the problem size, depending on the complexity of the algorithm used. Construct a table comparing $n, n \log_2(n), n^2, n^3, 2^n$, and $n!$ for $n = 2, 4, 8, 16, 64, 512$. (You may need to use something like Wolfram Alpha to compute some of these quantities).

Solution:

	2	4	8	16	64	512
n	2	4	8	16	64	512
$n \log_2(n)$	2	8	24	64	384	4608
n^2	4	16	64	256	4096	2.621×10^5
n^3	8	64	512	4096	2.621×10^5	1.342×10^8
2^n	4	16	256	65536	1.844×10^{19}	1.340×10^{154}
$n!$	2	24	40320	2.092×10^{13}	1.268×10^{89}	3.477×10^{1166}